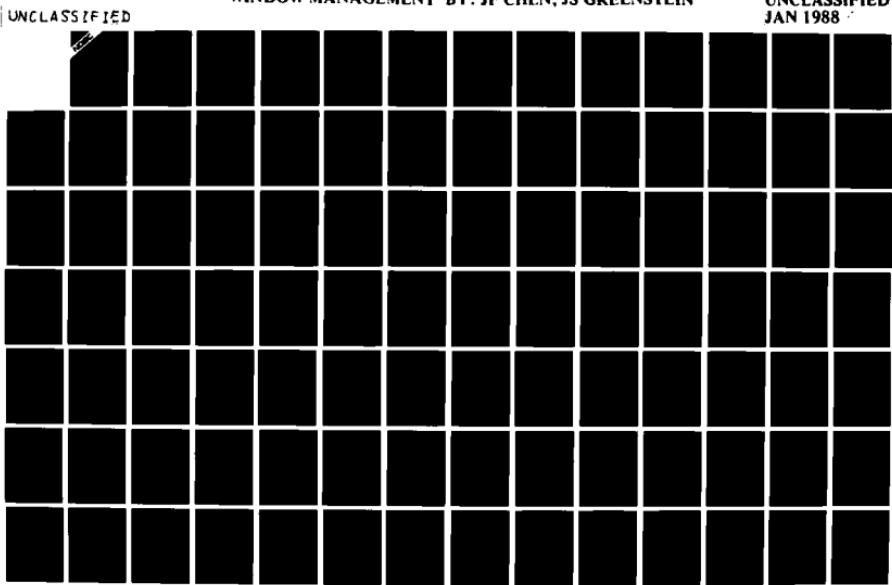


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LOCALIZED AND GLOBAL APPROACHES FOR TILED
WINDOW MANAGEMENT BY: JF CHEN, JS GREENSTEIN

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**Localized and Global
Approaches for Tiled
Window Management**

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Clemson University



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Window management systems provide a means to facilitate the arrangement of the displays of computer-based information. Some systems permit windows to overlap while others require that windows be stacked. Some systems will allow the user to determine the location and size of a new window while others require the user to locate and size the window. This research compared two approaches to window management: local or global. Local window systems -- the first, termed the localized approach -- creates space for a new or enlarged window by taking space from adjacent windows in the same row. Space is taken from nonadjacent windows only when adjacent windows do not have enough space to meet the needs of the new or enlarged window. The second approach, termed the global approach, creates space for a new or enlarged window by taking equal amounts of space from all the other windows in the same row. The objective of this research was to determine the relative efficiencies of these two approaches when used under each of four decision making situations. The results of a controlled, behavioral study of human performance suggest that one or the other of the two window management approaches is appropriate depending on: the amount of space available to display the data; the variability in the sizes of the data sets that are to be displayed; and the interdependence among these data sets.

100-100000

This letter will serve as my answer to the statement made by Mr. Joseph L. Rabinowitz, Chairman of the Board of Directors of the New York Stock Exchange, in his speech before the New York City Chamber of Commerce on January 22, 1938. Apparently he is referring to my letter of January 10, 1938, in which I informed him of the following fact: "The Department of Justice, the Federal Reserve Board, and the Comptroller of the Currency have all agreed that no member bank of the Federal Reserve System should accept or cash checks drawn on non-member banks." I would also like to thank you for your kind letter from Mr. P. G. Davis and Mr. C. W. Kimball for their valuable suggestions and assistance.

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CONTINUATION

The research has been concerned with the development of a window management system for the VLSI design environment. The interface design of the management system will allow the user to obtain window-based information. A window-based interface allows the user of a computer-generated display to simultaneously view multiple windows and multiple sources of information. It is anticipated that the window display and window manipulation functionality will allow the researcher to demonstrate concepts with a much more cost-effective variety of display hardware.

A window management system can be characterized by the set of characteristics it provides for window management. For example, some systems permit windows to overlap while others require that windows be tiled. Some systems automatically determine the location and size of a new window, while others require the user to locate and size the window.

The researcher compares two approaches to window management commonly used in window systems. The first, termed the localized approach, creates space for a new or enlarged window by taking space from adjacent windows in the same row. When the adjacent windows do not have enough space to meet the needs of the new or enlarged window, space is also taken from the next most adjacent windows. The second approach, termed the global approach, creates space for a new or enlarged window by taking space from all windows in the same row. The global approach is more efficient than the localized approach because it does not require the user to determine the position of the window.

of "Dolan 88" applications which are to different degrees of complexity, and four scenarios. A capture-recapture experimental study of Pectoral Shearwater distribution and abundance using the modified AT&T *g*-age application was conducted. The test was designed to permit comparison of the two approaches. The results of this study will be used to assess potential difficulties in the selection of a window management approach as a function of the amount of space available for information display, the variability of the application data sets that are to be displayed, and the characteristics of the those data sets.

CHAPTER 1.

LITERATURE REVIEW

The major factors issues derived from a literature review of window-based human-computer interfaces may be divided into seven categories:

1. Benefits

- 1. The benefits of using multiple windows
- 2. Window categories
- 3. Tiled windows vs. overlapping windows
- 4. User control vs. system control
- 5. Localized vs. global window management
- 6. Window management command syntax
- 7. Window management input devices

The Benefits of Using Multiple Windows

A window management system gives the user access to the full power of a multitasking operating system by allowing several interactive applications to run at once (Goodfellow, 1986). Without a window manager, the user is forced to run a single interactive application and is unable to use multitasking only for batch programs.

Window managers can serve at least seven functions (van Favel and Farrell, 1984):

- 1. They allow relatively rapid access to more information than would be accessible with a single frame of information using the same screen size.
- 2. They permit access to multiple sources of information.
- 3. They can be used to combine multiple sources of information.
- 4. They can be used to control multiple programs.

- 3. They can be used to provide the user easy access to information that may be useful in the near future. For example, pop-up windows containing menus of commands or a clock window giving the time can be displayed.
- 4. They serve as an indication of command context and active forms. When the display cursor is in a given window, commands take on a specific meaning or that window becomes active. This allows the command language to be simpler and more modular.
- 5. They can be used to display multiple representations for the same task. This can permit the user to perform a sequence of actions using the representation easiest to use at each point in the sequence.

Bury, Davies, and Darnell (1985) conducted an experiment to test the hypothesis that the user can gain benefits from multiple windows in a task which requires the display of supplemental information relevant to the user's primary task. Although users of the windowed environment spent additional time getting windows onto the screen and arranging them in a usable layout, their performance was more accurate than that of users in the non-windowed environment.

Bury et al. (1985) suggest that users can also benefit from multiple windows in the following tasks:

- 1. tasks in which users monitor system changes in a secondary window while performing a task in a primary window;
- 2. tasks in which specific locations within more than one application need to be specified

Window Categories

Card et al. (1986) stated that current designs for window management systems can be broken down into approximately four categories: simple "TTY" windows, time-multiplexed windows, space-multiplexed windows and non-homogeneous windows. Simple "TTY" windows imitate the mechanical flow of paper in a teletype machine. Time-multiplexed windows include scrollable windows and frame-at-a-time systems.

Figure 3 illustrates windows that are often used during the work process. The initial dimensions they employ within these groups, space-multiplexed windows may be further classified as independent, displaying separate information, or split, displaying subsets of closely related information. A one-dimensional window is a screen divided either vertically or horizontally into a number of separate parts. Two-dimensional windows are tiled windows and two-and-a-half dimensional windows are overlapping windows.

Non-homogeneous windows include those in which the level of the displayed information may change within or across windows. Icons, bifocal windows, fish-eye windows, and zooming windows are techniques for accomplishing this.

Icons are very small windows, generally represented on the screen by a small symbolic picture of some sort. An icon may be selected and expanded into a full-size window. Icons are a means for keeping reminders of a large amount of information on the screen without taking up much space.

Bifocal windows include those windows in which hierarchical information is displayed in full detail in the center. Related information is displayed on the periphery in just enough detail to recognize it. Thus, the user always has detailed display of some item of interest and no detailed display of contextually related items.

In optical fish-eye windows, information is compressed in the window like the image of a convex mirror. In logical fish-eye windows, information detail displayed on the screen may be reduced according to the logical distance from some focal point. For example, a program listing may be completely displayed to one arm of a conditional

statement of interest in the original statement, and then the user is invited to each of the other arms of the conditional statement.

In zooming windows, data in a window or the window itself can be made larger or smaller in the manner of a zooming camera lens. This effect may involve changing type fonts, suppressing parts of the diagram, or even distorting parts that are most important.

Cohen, Smith, and Iverson (1986) reported that four different kinds of tiled window organizations have been built:

1. Single column--Windows are lined up vertically in a single column.
2. Multicolumn--The screen is divided into multiple columns, and windows are lined up vertically in each column.
3. Hierarchical--The screen is divided either vertically or horizontally into partitions, and the partition is then recursively divided into subpartitions. Windows correspond to the final set of undivided partitions. Each partition corresponds to a node in a hierarchy, with windows at the leaves.
4. Nonhierarchical--Any rectangular tiling can appear in the system.

Tiled Windows vs. Overlapping Windows

A tiled window system is defined as one in which any open window is always fully visible; windows are not allowed to overlap. An overlapping window system is defined as one in which the user can manage a window's location and size in any way desired.

Goodfellow (1986) noted that tiled windows and overlapping windows both have advantages. The advantages of tiled windows include

1. Since no window is obscured, the user never has to rearrange the screen to see something that has been buried accidentally, as is common with the overlapping model.
2. The screen is better utilized since it is completely covered with windows.
3. The architecture is easy to implement.

- The window manager is responsible for managing windows. The window manager does not have to be concerned with the contents of the windows.

The advantages of overlapped windows are:

1. A new window is created on top of the previous windows. This approach is more intuitive than tiled windows.
2. With the tiled model, each window has its own icon. These icons are iconized when a new window is created. The user must then rearrange these icons if he wants to see them. In the overlapped model, the user can identify icons with their titles. The user can also overlap windows with the overlapping model. If a window is not needed it can be moved out or iconized.
3. Overlapping windows permit the user to place a window wherever he wants, unconstrained by the position and size of other windows.
4. More windows can be displayed at once using overlapping windows.

Cohen et al. (1986) reported that users of the desktop model, which allows windows to overlap each other like sheets of paper piled up on a physical desk, pay a price for its flexibility, especially when large numbers of windows need to be accommodated. The desktop can become messy, forcing the user to constantly locate and rearrange windows. Overlapping windows may be most useful for systems using small screens or for applications that use only a few windows over a long period of time. Compared with overlapping windows, tiled systems which allow the user to control the tiling are safe. Users do not need to worry about what has been buried accidentally but can require enough work of the user to become frustrating. A tiled system which automatically retiles the screen based on the user's action can relieve the user of window management tasks, but it may locate and size windows inappropriately requiring the user to adjust the layout manually. It is possible to place constraints on the reordering and tiling of the windows.

however. The user should feel restricted by the constraints imposed by the constraints help to ensure that the right windows are in the screen in the right place with the right size. Tiled windows may be especially useful for applications that generate large numbers of small, short-lived windows.

Bly and Rosenberg (1986) conducted an experiment to compare tiled windows with overlapping windows and reported that tasks which require much more window manipulation (tasks in which users must frequently arrange the windows and make the necessary text visible within the borders of the windows, either by scrolling or changing the window sizes can be carried out more quickly using overlapping windows, while tasks which require little window manipulation can be carried out more quickly with tiled windows. If users are inexperienced in the use of overlapping windows, tiled windows are better for both kinds of tasks. The authors suggest the possibility that for a user equally proficient with overlapping and tiled window systems, tiled windows may be better in both situations. Paradoxically, however, 19 of the 22 subjects studied preferred overlapping windows to tiled windows.

User Control vs. System Control

Bury et al. (1985) reported that one of the most significant ways in which window management systems differ is in whether the windows are under user or system control. Some of the most effective uses of multiple windows have been in systems in which the configuration of the windows is entirely under system control. System controlled windows often negate all the advantages of windows, while enabling management of their display instances. In many cases, the window management systems will be intelligent enough to decide the best location and size for a window.

allowing the user to choose where to open it. Some systems give the user control they allow depending on the type of window. Menu, message and help windows are often less modifiable than work windows. Some systems give the user no control over the placement of menu windows. Other systems allow the user to specify the location where the window will be opened but nothing else. Still others allow the user to move the window around after it has been opened. In most cases once a selection from the menu list is made the window is automatically removed. Still other systems make a distinction between types of work windows, giving certain types of work windows full customizability while limiting the function of others.

Localized and Global Approaches for Window Management

Cohen et al. (1986) stated that tiled window management systems differ widely in their approaches to resizing and opening a window. When a window is resized in Andrew and RTL/RTL (two of the several different window management system implementations that have been reported in the literature), effects are local and manual (only the adjacent window is given or gives up space and the amount of space given or taken is controlled by the user).³ In Cedar and RTL/CRTL, effects are global and automatic (all the other windows are given or give up space with the amounts of space given or taken controlled by the system). In Gosling's EMACS system the effect is as local as possible. If the adjacent window does not have enough space, space is taken from the next most adjacent window. In the RTL/RTL system, after specifying the window to be opened, the user locates it at the edge of a tile and then pulls on the edge to make space for it. Both Cedar and Microsoft Windows support

that takes place when three other windows proportionately. Microsoft Windows allows the creation of window Microsoft Windows also supports the automatic placement techniques. In one, the window is simply located at the edge of a tile, and space for the new window is taken away equally from the windows on both sides of it. Microsoft Windows also allows the user to locate a window on top of an existing window completely replacing that window. Andrew supports both manual and automatic placement. For automatic placement, it finds a window large enough to be split so that even after space for the new window is taken away from it, it is still larger than its size constraints require. In manual placement the specified location is used to find a location in the window hierarchy, and this determines the actual placement.

Window management systems also differ greatly in their approaches to closing a window as a result of resizing or opening another window. In systems where resizing has local effects, enlarging a window takes space from an adjacent window. When the adjacent window becomes too small, either enlargement is prohibited (e.g., Gosling's EMACS) or the adjacent window is automatically closed (e.g., Andrew and RPL). RPL Andrew may also close a window when another window is opened. Neither Cedar nor Microsoft Windows automatically closes windows. Cedar will not allow a window to be enlarged if this will force all the other windows in its column to be shrunk below their minimums, but it will open a new window, even if this shrinks the remaining windows below their minimums. Microsoft Windows will not open a window if there is no space available for it. RPL/RPL has a variety of options for treating automatic closure. If the user specifies explicit closure, the system

such a system will be able to distinguish windows by placement, size and alignment and will be able to determine if the user has attempted to close the user's chosen window. If the user has been permitted automatic window closure in the CRTG then the windows to be closed may be required to be local so that only windows adjacent to the opened window will be closed. Alternatively windows will be closed on the basis of size and priority.

Window Management Command Syntax

Bury et al. (1985) stated that different systems can have semantically identical command interfaces, but be radically different in their command syntax. Some systems adopt an action-object command structure while other systems adopt an object-action command structure. It is easier to construct effective user prompts for a system which uses an action-object syntax structure. This is because effective prompts require knowledge of the user's intentions and the user's intentions can be deduced to a greater extent with knowledge of the action than with knowledge of the object. On the other hand, the object-action syntax structure usually results in a system with fewer modes and this is generally regarded as desirable.

Window Management Input Device

Bury et al. (1985) noted that different systems might require users to take the following actions to issue commands:

1. Point at and select actions and objects with a mouse
2. Point at and select with cursor keys
3. Type a command - for instance "Move window to top left"
4. Issue command via function keys
5. Issue command via voice

The interface of the system and the user interface are often tightly interrelated. The user interface is represented to the user end-user with words or at least partly with graphic icons. The basic syntax of commands might be the same in all of these systems but the command mechanisms might be very different.

PARTER III

RESEARCH OBJECTIVE AND HYPOTHESES

The literature review makes it clear that the window management system provides a very flexible environment for the display of computer based information. Although many window management systems have been developed and a number of researchers are pursuing the development of even more powerful systems (Hopgood, Duce, Fielding, Robinson, and Williams 1986), only two empirical investigations of human performance with such systems appear to have been reported in the open literature (Sly and Rosenberg 1986), investigating tiled windows vs. overlapping windows, and Bury et al. (1985), investigating the benefits of multiple windows. Since the principal reason for using window management systems is to gain the benefits of a better human-computer interface, it is important for system designers to know, for example, in which situations the localized approach for resizing and opening tiled windows is better than the global approach and in which situations system control of window management is better than user control. Such information would permit designers to implement the window management system appropriate for a given situation.

Because users of tiled window management systems may need to spend substantial amounts of time arranging windows into configurations suitable to perform tasks particularly if windows become too small to fully display the information associated with them, it is worth investigating the effects of localized and global approaches for tiled window management on overall task performance. The objective of this research is to

determine the relative efficiency of these two approaches under various decision making scenarios.

As noted in the preceding literature review, there are two categories of systems differ in their approaches to opening and resizing a new window. In some systems the effect of opening and resizing a new window on the existing windows is localized, with space created for a new or enlarged window by taking space from adjacent windows. When the adjacent windows do not have enough space to meet the needs of the new or enlarged window, space is also taken from the next most adjacent windows. In other systems the effect is global, with space created for a new or enlarged window by taking space from all the other windows in the same row or column of the display. The effects of these two approaches on the layout of a display are illustrated in Figures 1, 2, and 3.

Three hypotheses are proposed regarding the effects of these two approaches on the amount of time the user must devote to window management. In stating these hypotheses, it is assumed that a maximum of M rows of windows are permitted on the display. The mean number of data sets that can be displayed in their entirety within the spatial confines of each of these M rows is denoted by N . Thus, on average, the total area available on the video display is sufficient to permit the simultaneous display of $M \times N$ data sets in their entirety. When less than $M \times N$ data sets are displayed on the video display at once, it is likely that the window allocated to each of the displayed data sets will be large enough to display the full contents of the data set. When more than $M \times N$ data sets are displayed on the video display at once, it becomes more likely that the windows allocated to one or more of the displayed data

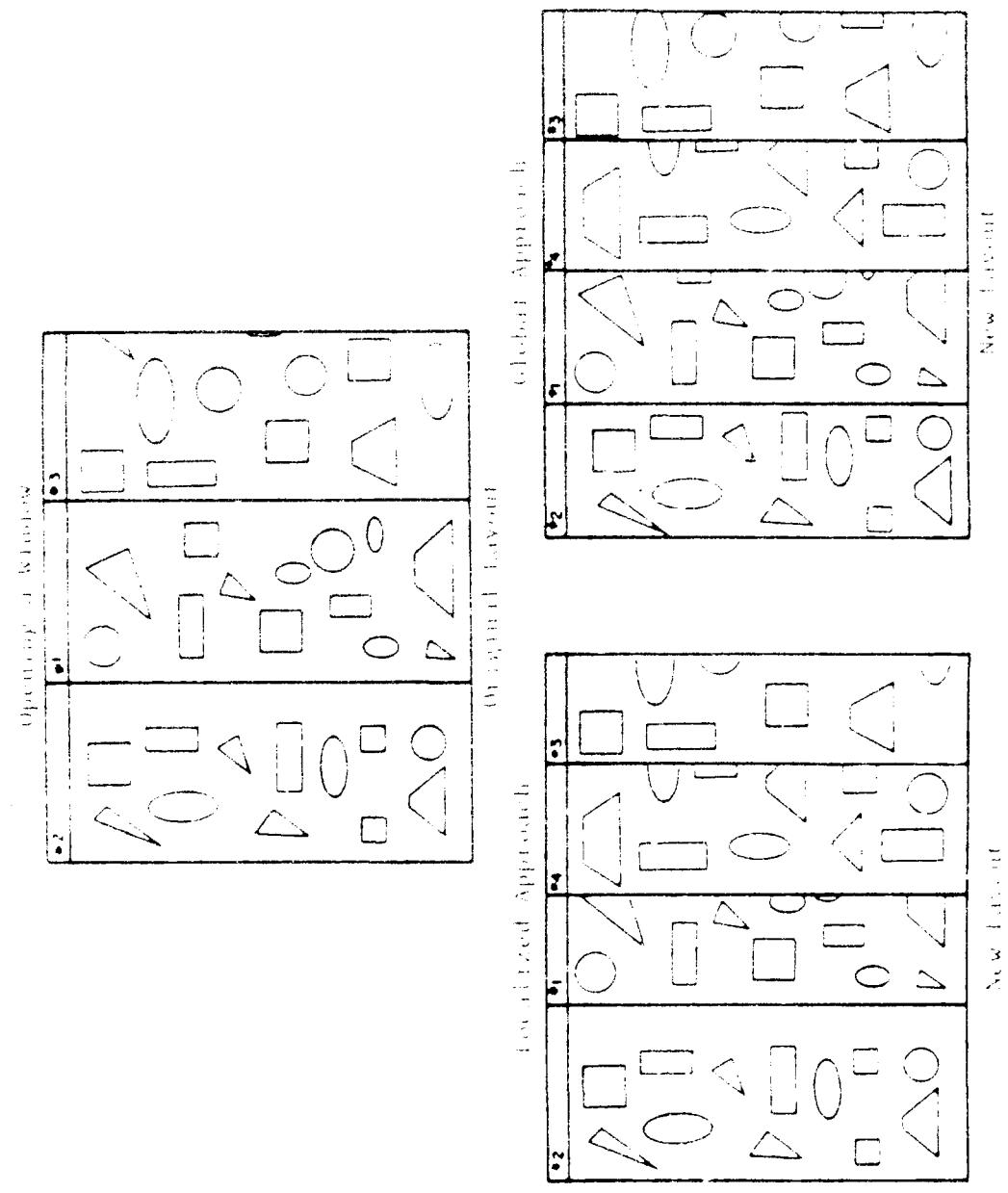


FIGURE 1 - Localized Approach vs. Global Approach in Opening a Window (3)

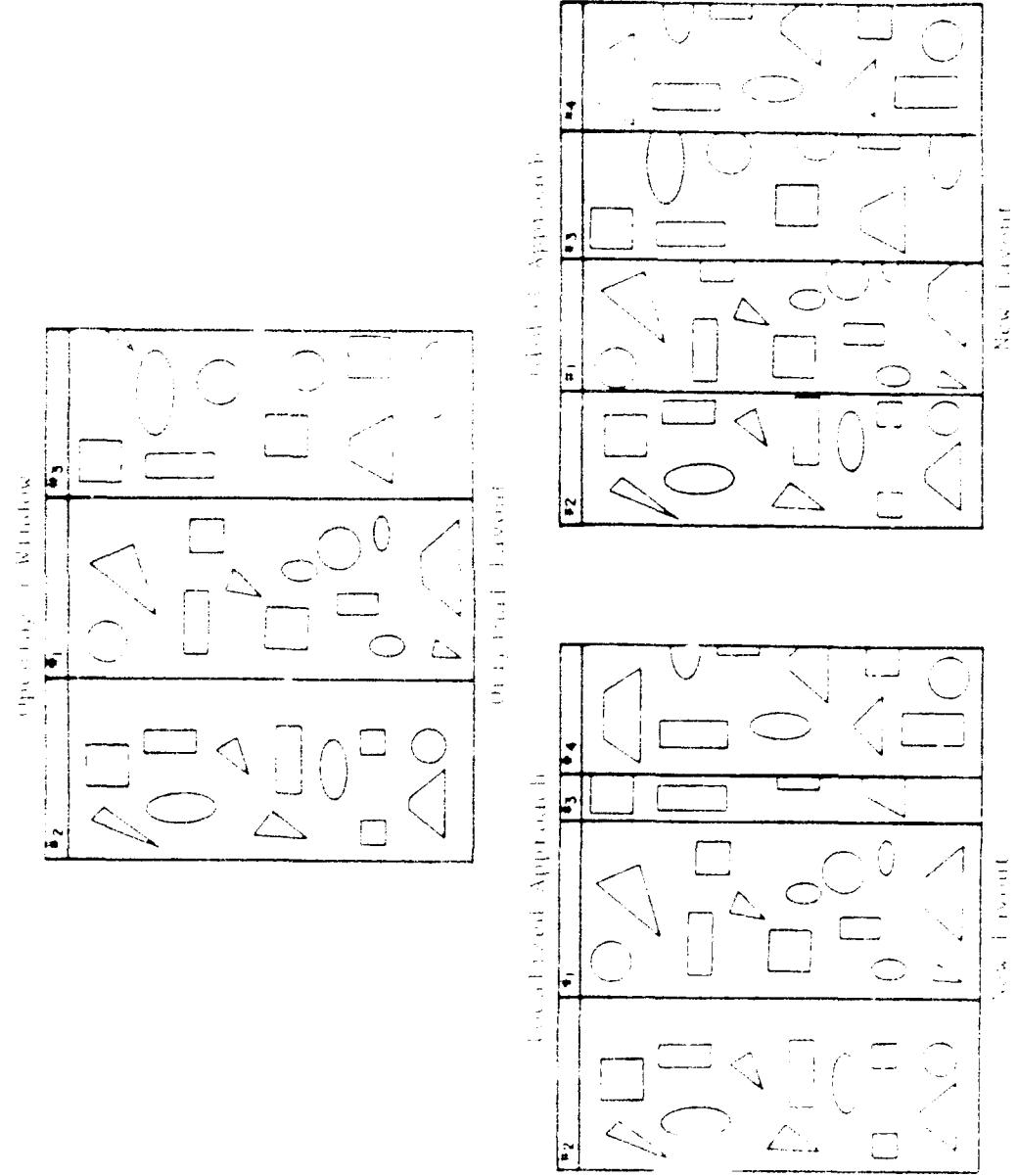


Figure 3: Localised Approach vs Global Approach in Opening a Window (b)

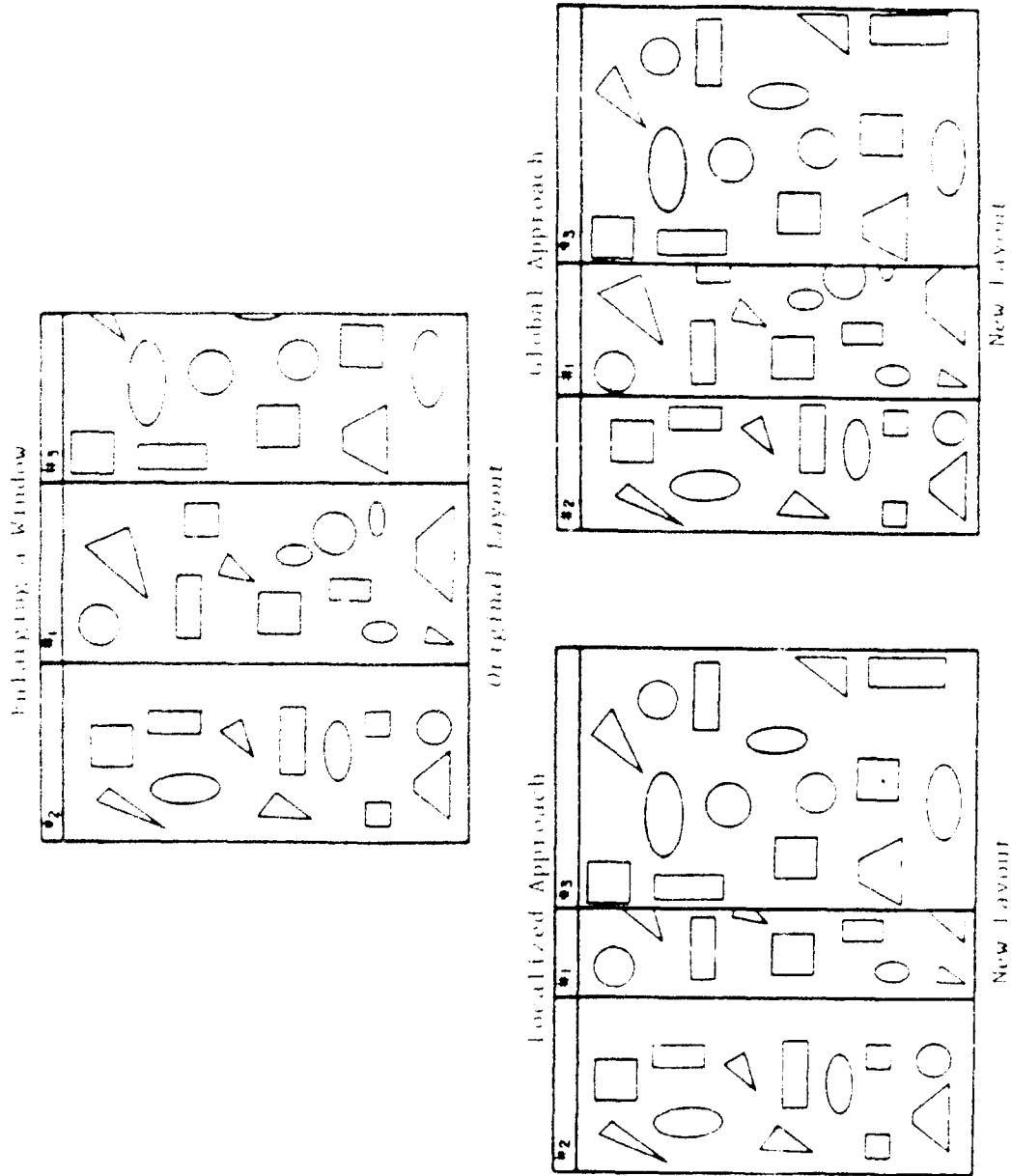


Figure 3 Localized Approach vs. Global Approach in Enlarging a Window

sets will result in the largest amount of display time spent in the data set.

Hypothesis 1

If there are less than $M \times N$ windows on the display and the sizes of the different data sets required to perform a task are approximately equal, then a system employing a global approach to window management will be more efficient than an otherwise equivalent system employing a localized approach. (Figure 4 illustrates two data sets that are of equal size.)

The basis for this hypothesis is as follows: When there are less than $M \times N$ windows on the display and the data sets displayed in these windows are approximately equal in size, it is likely that each of the windows contains space in excess of that required to display the full contents of the associated data set. Thus, if an existing window needs to be enlarged or a new window needs to be opened, the small amount of space that will be taken from each window to accomplish this operation under the global approach may leave these windows sufficiently large to continue display of the full associated data sets. Under the localized approach, however, space will be taken from only one or two windows. The large amount of space that each of these windows must give up may leave these windows too small to continue display of the full associated data set. Thus enlarging an existing window or opening a new window may have to be followed by additional window enlarging operations.

Hypothesis 2

If there are less than $M \times N$ windows on the display, the sizes of the different data sets required to perform a task vary greatly and the

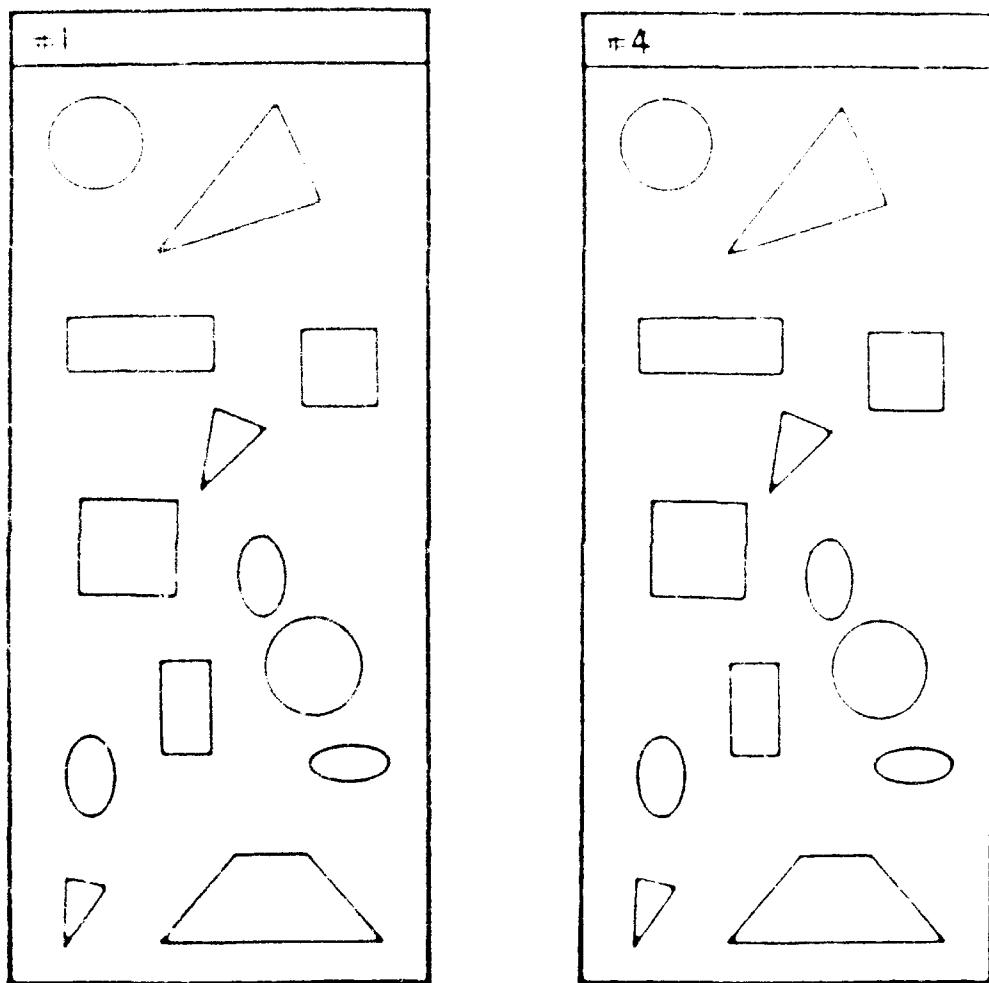


Fig. 10.4 - New Variability in Data Sets Given

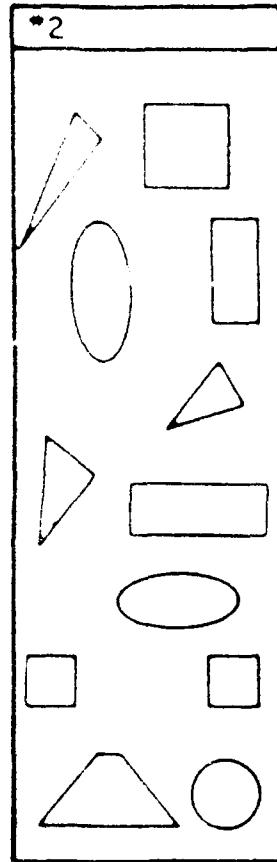
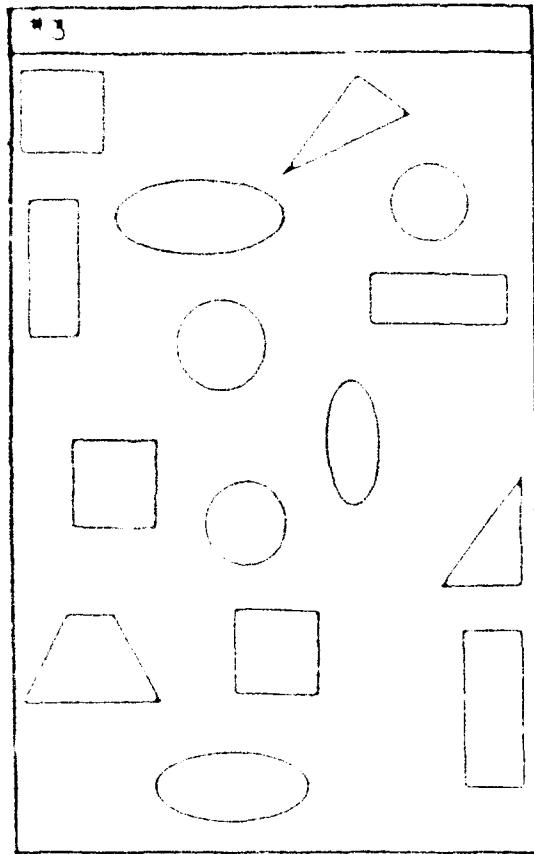
data sets are highly interdependent. Then, a system employing a localized approach to window management will be more efficient than an otherwise equivalent system employing a global approach. Figure 4 illustrates two data sets that vary greatly in size.

The basis for this hypothesis is as follows: When there are less than $M \times N$ windows on the display and the data sets displayed in these windows vary greatly in size, it is likely that some windows contain a great deal more space than that required to display their contents, while other windows do not contain much excess space. If the data sets required to perform a task are highly interdependent, the full contents of several of these data sets may need to be displayed simultaneously to permit performance of the task. If an existing window needs to be enlarged or a new window needs to be opened, the localized approach permits the user to take space only from those windows with sufficient excess space to meet the needs of the operation. Under the global approach, more windows would be affected by the operation, increasing the likelihood that windows without much excess space would be reduced below the size needed to view their full contents. This, in turn, would increase the likelihood that additional enlarging operations will be required.

Hypothesis 2

If there are $M \times N$ or more windows on the display and the data sets required to perform a task are highly interdependent, then a system employing a localized approach to window management will be more efficient than an otherwise equivalent system employing a global approach.

The basis for this hypothesis is as follows: When there are $M \times N$ or more windows on the display at once, it is likely that the additional



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to will have to wait for a window with the size required to display the full contents of the associated data set. If the data sets required to perform a task are highly interdependent, the full contents of several of these data sets may have to be displayed simultaneously to permit performance of the task. Thus, if an existing window needs to be enlarged or a new window needs to be opened, the localized approach will reduce only one or two windows below the size needed to view their full contents. Only if the data sets contained in these windows are necessary to complete the task will any additional enlarging operations be necessary. Under the global approach, however, several windows will be reduced below the size needed to view their full contents, increasing the likelihood that additional enlarging operations will be required.

CHAPTER IV
EXPERIMENTAL PLAN

An experiment was designed to test the relative efficiencies of the localized and global approaches to tiled window management when used under each of four decision making scenarios. In this experiment the maximum number of rows permitted on the display (M) was fixed at two, while the mean number of data sets that can be fully displayed within each of these rows (N) was fixed at five.

Subjects

Eight Clemson University students participated in this study. Volunteers were screened for corrected 20/20 vision at a distance of 67 cm. They were paid \$30 for the time spent in the experiment (7 to 8 hours).

Apparatus

The hardware configuration consisted of a Tandy 3000 HD personal computer with keyboard interfaced to a Tandy VM-3 monochrome text monitor and an NEC Multisync JC-1401P3A color graphics monitor. Software was written to produce a tiled window system manager capable of opening, enlarging, and reopening windows in each of at most two rows using either of the localized or global approaches to window management. The software was also designed to permit automatic closing of existing windows when opening, enlarging, or reopening another window reduces the amount of space available to the existing window below a specified minimum.

The task environment consisted of a number of data sets. Each participant saw a variety of two-dimensional graphical objects (e.g., circles, ellipses, squares, rectangles, parallelograms, and trapezoids). Participants in the study were asked to determine specific facts from these data sets using the window management system (e.g., "What is the total number of squares contained in data sets b, c, and d?"). The window management system could be used to open or reopen a specified window at a specified location. It could also be used to enlarge a specified window through movement of the left or right window boundaries to specified locations. All textual input and output (questions to be answered by the participant, window management commands entered by the participant, feedback directed to the participant, and answers entered by the participant) were directed to the monochrome text monitor (see Figure 6). All data sets were displayed in a graphical windowed environment on the graphics monitor in white on a black background. Figure 7 illustrates this environment. Asterisks denote the endpoints of data sets. The DATA SET MENU contains icons of data sets that have yet to be displayed. The HIDDEN DATA SETS menu contains icons of data sets that have been removed from the display to make room for other data sets.

Independent Variables

Three independent variables were investigated. The first window management approach was tested at two levels: localized and global. The second and third independent variables together characterized four different decision-making environments in which windowing systems might be employed. The second independent variable, variability of data set size, was tested at two levels: low and high. In the low variability

||||||||||||||||||||||||||||
| ENTER 'd' TO DISPLAY A DATA SET |
| ENTER 'e' TO ENLARGE A WINDOW |
| ENTER 'r' TO REDISPLAY A DATA SET |
| ENTER 'a' TO GIVE AN ANSWER |
||||||||||||||||||||||||

QUESTION 26

What is the total number of circles and squares contained in data sets h and k?

>a

GIVE YOUR ANSWER

>7

This answer is correct. Enter any key to continue

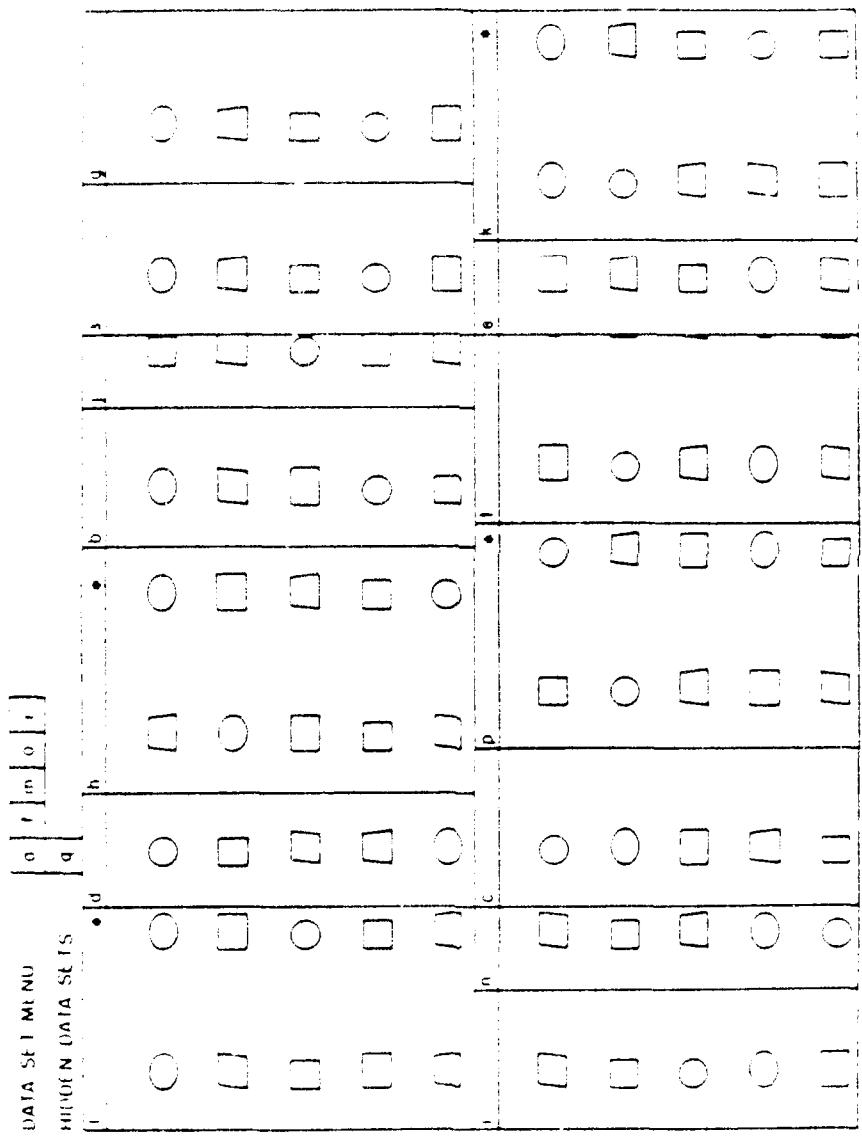


Figure 3: Windowed Data Sets Displayed on the Graphical Display

task all the data sets required equal amounts of space to display their contents. This amount of space was an area with height equal to one-half the display height and width equal to one-fifth the display width. In the high variability case the mean amount of space required to display the full contents of a data set was the same as in the low variability case, but some data sets required little space for complete display, while others required much larger amounts of space. Data set widths were normally distributed with a standard deviation equal to one-fourth of the mean data set width. The third independent variable, interdependency of data sets, was tested at two levels: low and high. In the low interdependency case participants were required to retrieve information from two data sets at a time to determine facts (e.g., What is the difference between the total number of trapezoids and rectangles contained in data set g and the total number of circles and parallelograms contained in data set k?). In the high interdependency case two to four data sets needed to be accessed to determine facts (e.g., What is the total number of squares contained in data sets e, o, r, and t?). The task sets employed in the low and high data set interdependency treatment conditions are included as Appendix F.

Experimental Design

There were eight treatment conditions resulting from the crossing of two window management approaches, two levels of data set size variability, and two levels of data set interdependency. A completely within-subject experimental design was used (see Figure 8).

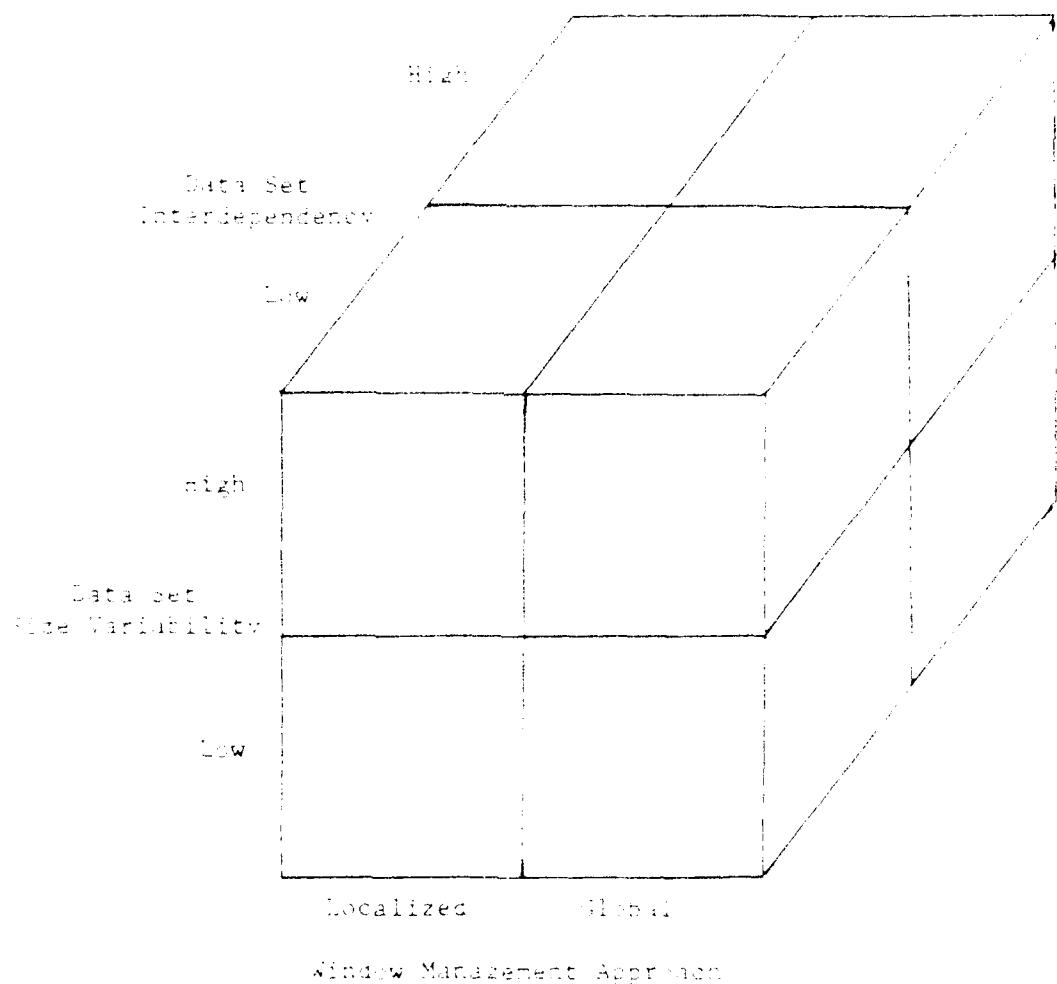


Figure 3. Experimental Design

Procedure

Two window management systems employing the localized and global approaches respectively were used in this study. Subjects used these two window management systems to complete the eight task blocks resulting from the crossing of two levels of window management approach, data set size variability, and data set interdependency. The order of administration of these eight task blocks was counterbalanced across subjects. Each of these eight task blocks required the subject to extract information contained in 20 data sets to perform 36 subtasks (i.e., 36 questions requiring the subject to use the window management system to determine specific facts from the 20 data sets). The 36 subtasks in each treatment condition can be divided into two phases.

The first phase included the first 18 subtasks. To complete the first subtask the subject had to open two windows to accommodate two data sets from which the subject needed information. In each of the 3rd, 5th, 7th, 9th, 11th, 13th, 15th, and 17th subtasks, the subject had to open an additional window to accommodate a new data set from which the subject needed information. To complete these and the other subtasks in this phase, the subject might also need to enlarge or reopen windows. Enlarging would be necessary if any of the windows containing data sets necessary to complete a subtask were too narrow to display the full contents of the associated data set. Reopening would be necessary if any of the windows containing data sets necessary to complete a subtask were automatically closed. The first 18 subtasks were categorized as a group because ten windows had to be opened to complete all subtasks at this phase. This was the average number of data sets that could be

displayed on the screen simultaneously. The subject had to switch his display.

The second phase was called the "switching" phase. To complete these subtasks, the subject had to open two windows to accommodate two new data sets from which the subject needed information. It consisted of the 5th, 6th, 7th, 8th, 11th, 12th, 13th, 10th, and 17th subtasks in this phase. The subject had to open an additional window to accommodate a new data set from which the subject needed information. To complete these and the other subtasks in this phase, the subject might need to enlarge a window if a window containing a data set necessary to complete a subtask was too narrow to display the full contents of the data set. The subject might also need to reopen a window if a window containing a data set necessary to complete a subtask was automatically closed.

Each subject was tested over four days. On the first day, the subject signed an informed consent form, read the instructions, and was given tutorials on the use of the window management system and on the nature of the tasks (Appendices A, B, and C). Each subject completed two task blocks per day. Before each of the task blocks, the subject completed a training block to gain familiarity with the commands and features of the window management system he would be using and with the nature of the tasks he would be expected to perform. The subject then performed the actual task block. Next, during a rest break, the subject completed portions of a questionnaire regarding his satisfaction with the window management approach tested in the preceding treatment (indicated Appendix J). At the end of the study, the subject completed a second questionnaire concerning his overall satisfaction with respect to the use of the new management approach (Appendix K).

Dependent Variables

The task completion time, number of window management operations (i.e., openings, enlargings, and reopenings), and number of incorrect solutions were recorded for each of the eight treatment conditions. Task completion time was divided into two primary components: window management time and task solution time.

Window management time included any time the subject spent arranging the display screen into a configuration that was suitable to complete the subtask. Window management time itself was composed of:

1. opening time which included the time spent opening windows;
2. enlarging time which included the time spent enlarging windows;
3. reopening time which included the time spent reopening windows.

Task solution time was the time the subject devoted to determining the specific facts requested, once he had arranged the windows into a configuration suitable to perform this task.

To permit testing of the previously stated research hypotheses, the dependent measures were categorized as follows:

1. Measures based on the performance of subtasks 1 through 18. Subtasks 1 through 18 were performed under conditions in which the number of windows on the display was less than the mean number of data sets that could be fully displayed within the area available on the video display ($M \times N = 2 \times 5 = 10$).
2. Measures based on the performance of subtasks 19 through 36. These subtasks were performed under conditions in which the number of windows on the display was equal to or greater than the mean number of data sets that could be fully displayed within the area available on the video display.
3. Measures based on the overall performance of all 36 subtasks. These measures provide an overall indication of the relative efficiencies of the two window management approaches for the tested task sets.

At the end of each treatment condition portions of a subjective rating scale questionnaire were completed by each participant to permit

• Evaluation of three window management methods applied to decision making tasks
perceived ease-of-learning and satisfaction. At the conclusion of the study, the subject completed a second questionnaire ascertaining his overall satisfaction and preference for the two window management approaches.

Analysis

A three-factor within-subject analysis of variance was performed to determine which window management approach provided the most effective human-computer interface under each of the four decision making scenarios and during each of the two phases tested (less than 10 displayed data sets and 10 or more displayed data sets). Analytical comparisons were performed, where appropriate, to determine the locus of any significant interactions among the independent variables that were identified by the analysis of variance. An analysis of variance summary table is presented in Table I.

Table 1 Analysis of Variance Summary Table for a Three-factor Within-Subject Design

Source	df	SS	MS	F
<u>Between Subjects</u>				
S	(n-1)	SS_S	MS_S	
<u>Within Subject</u>				
A	(a-1)	SS_A	MS_A	$MS_A / MS_{A \times S}$
AxS	(a-1)(n-1)	SS_{AxS}	MS_{AxS}	
B	(b-1)	SS_B	MS_B	$MS_B / MS_{B \times S}$
BxS	(b-1)(n-1)	SS_{BxS}	MS_{BxS}	
C	(c-1)	SS_C	MS_C	$MS_C / MS_{C \times S}$
CxS	(c-1)(n-1)	SS_{CxS}	MS_{CxS}	
AxB	(a-1)(b-1)	SS_{AxB}	MS_{AxB}	$MS_{AxB} / MS_{AxB \times S}$
AxBxS	(a-1)(b-1)(n-1)	SS_{AxBxS}	MS_{AxBxS}	
AxC	(a-1)(c-1)	SS_{AxC}	MS_{AxC}	$MS_{AxC} / MS_{AxC \times S}$
AxCxS	(a-1)(c-1)(n-1)	SS_{AxCxS}	MS_{AxCxS}	
BxC	(b-1)(c-1)	SS_{BxC}	MS_{BxC}	$MS_{BxC} / MS_{BxC \times S}$
BxCxS	(b-1)(c-1)(n-1)	SS_{BxCxS}	MS_{BxCxS}	
AxBxC	(a-1)(b-1)(c-1)	SS_{AxBxC}	MS_{AxBxC}	$MS_{AxBxC} / MS_{AxBxC \times S}$
AxBxCxS	(a-1)(b-1)(c-1) (n-1)	$SS_{AxBxCxS}$	$MS_{AxBxCxS}$	
Total	abc(n-1)	SS_{Total}		

Factor A Window Management Approach

Factor B Data Set Size Variability

Factor C Data Set Interdependence

a=3, b=2, c=2, n=8

CHAPTER II

RESULTS

Five dependent measures were employed: window management time, number of window management operations, task solution time, number of incorrect solutions, and total task completion time. Each of these was analyzed separately for the initial portion of the task block when less than 10 data sets had been displayed on the screen, the remainder of the task block when 10 or more data sets had been displayed on the screen, and the entire task block.

Window Management Time and Number of Window Management Operations

Less Than 10 Displayed Data Sets

Tables II and III contain the ANOVA summary tables for window management time and number of window management operations, respectively. The Window Management Approach (A) x Data Set Size Variability (Dv) x Data Set Interdependency (Di) interaction is significant for both measures. Figures 9 and 10 depict this interaction for management time and number of management operations, respectively. Simple-effects F-tests (Tables IV and V) indicate that the simple A x Di interaction is significant for both measures when data set size variability is high. Multiple t-tests indicate that when data set size variability is high, the global approach results in significantly less window management time and fewer management operations than the localized approach if data set interdependency is low, while it results in more management time and management operations than the localized approach if data set

Table II. ANOVA Summary Table, showing main effects of the independent variables across the three displayed data sets.

Source	df	SS	F	p
<u>Between</u>				
Subjects x Sub	1	7628.10		
<u>Within</u>				
Approach x A	1	3274.21	36.02	0.0005
A x Sub	1	1024.99		
Data Set Size				
Variability x Dv	1	5296.75	43.42	0.0003
Dv x Sub	1	853.98		
Data Set Inter-dependency (Di)	1	1505.15	29.24	0.0010
Di x Sub	1	360.36		
A x Dv	1	6467.18	64.13	0.0001
A x Dv x Sub	1	700.95		
A x Di	1	106.12	0.67	0.4403
A x Di x Sub	1	1110.05		
Dv x Di	1	75.06	0.23	0.6430
Dv x Di x Sub	1	2240.81		
A x Dv x Di	1	1358.94	9.53	0.0176
A x Dv x Di x Sub	1	998.16		
Total	62			

Table 10. ANOVA Results: Difference in Number of Window Management Operations Less Than 10 Displayed Data Sets

Subject	df	SS	F	p
<u>Between</u>				
Subjects (S)	2	10.20	1.33	.30
<u>Within</u>				
Approach (A)	1	110.56	20.46	0.0001
A x Sub	2	10.44		
Data Set Size (D)	1	126.56	23.87	0.0001
Dv x Sub	2	9.44		
Data Set Inter-Dependency (Di)	1	26.43	16.31	0.00148
Di x Sub	2	13.75		
A x Dv	1	132.26	61.76	0.0001
A x Dv x Sub	2	14.72		
A x Di	1	3.56	0.32	0.5912
A x Di x Sub	2	1.44		
Dv x Di	1	1.56	0.78	0.4051
Dv x Di x Sub	2	13.94		
A x Dv x Di	1	15.06	17.20	0.0003
A x Dv x Di x Sub	2	6.50		
Total	63			

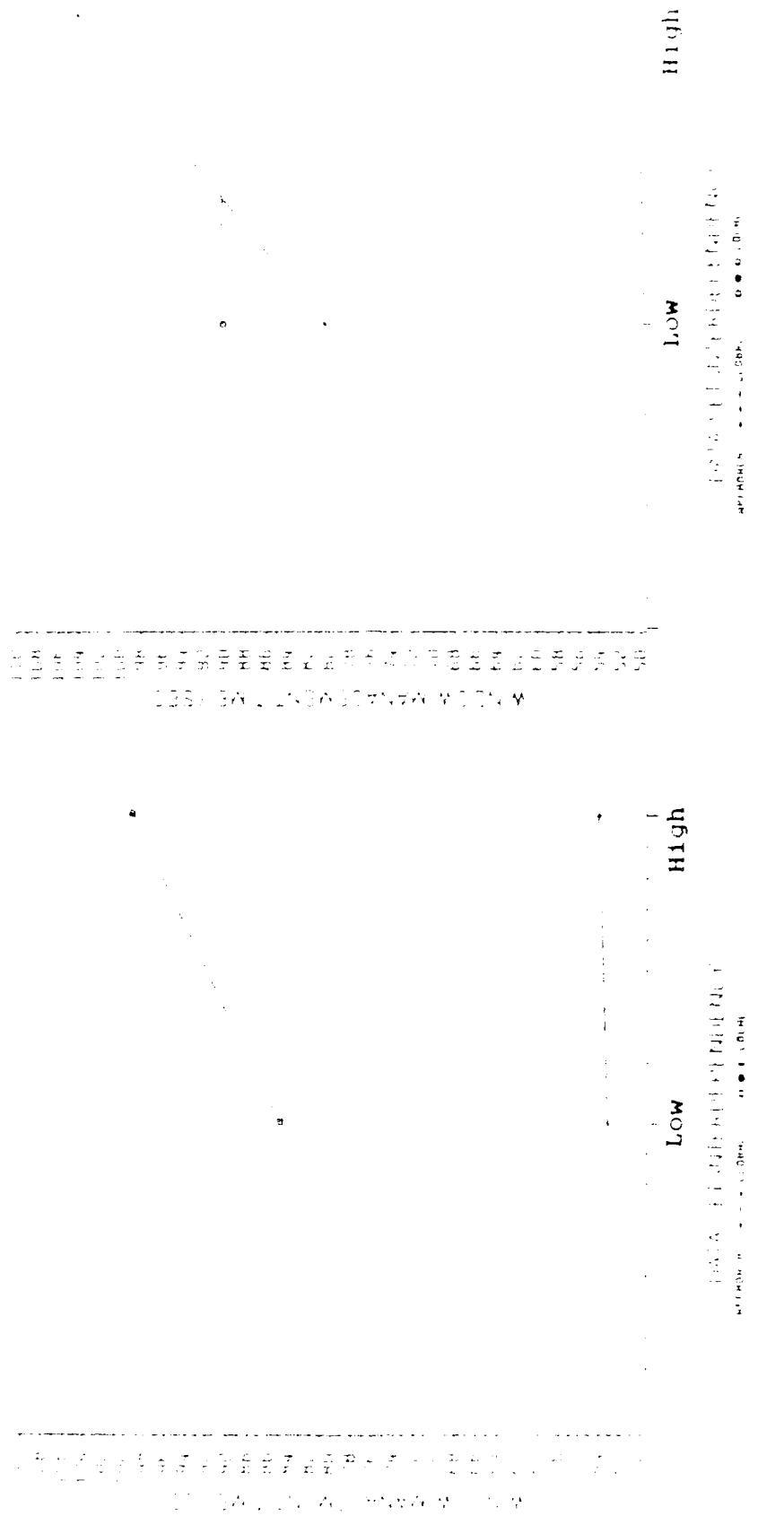


Figure 9. A x D₂ bi interaction for Window Management Time (less than 10 displaced data sets)

a. High Data Set Size Variability
b. High Data Set Size Variability

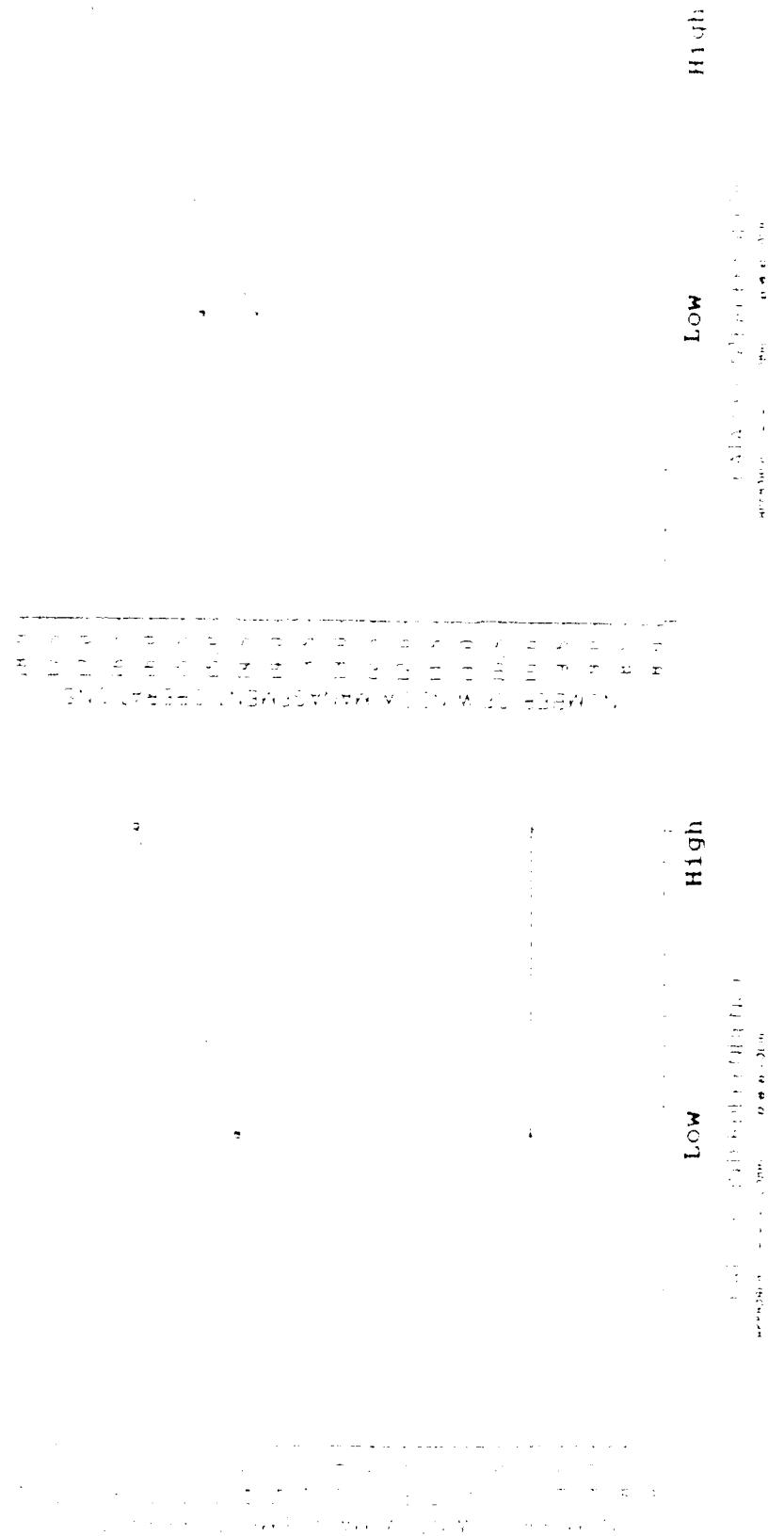


Figure 10. A \times D₂ \times W₁ interaction for Number of Window Management Operations (class Theta_10 displayed Data Sets)

a. Low Data Set Size Variability b. High Data Set Size Variability

Table II. Summary Table of the ANOVA Results for the Time Management Time (less than 100 displayed data set).

Source	S1	S2	F	P
Low data set size variability				
Approach (A)				
Data set inter-dependency (D1)	1	11680.47	130.24	0.0001
A x D1	1	348.02	3.48.02	0.2534
High data set size variability				
Approach (A)				
Data set inter-dependency (D1)	1	1126.10	8.62	0.0018
A x D1	1	1112.27	16.46	0.0048

Table 4. Summary statistics of the results of the experiments on clustered window management applications using different dispersed data sets.

Source	<i>n</i>	<i>M</i>	<i>S</i>	<i>p</i>
Low data set size variability				
Approach A	1	206.53	86.45	0.1061
Data set inter-dependency ('Di)	1	5.28	2.74	0.1777
A x Di	1	5.28	2.74	0.1777
High data set size variability				
Approach A	1	0.78	0.92	0.3701
Data set inter-dependency ('Di)	1	16.53	10.31	0.0148
A x Di	1	11.28	31.99	0.0008

interdependency is high (Table III). This significant simple A x Di interaction suggests that when there are less than ten data sets on the display and data set size variability is high, performance with the global approach is sensitive to changes in data set interdependency, while performance with the localized approach is insensitive.

The simple-effects F-tests on the A x Dv x Di interaction reveal a significant simple main effect of window management approach on both dependent measures when data set size variability is low (see Tables IV and V). This effect indicates that when data set size variability is low, the global approach requires less window management time and fewer window management operations (33.67 s and 10.00 operations) than the localized approach (91.89 s and 15.44 operations) regardless of the level of data set interdependency.

The simple-effects F-tests on the A x Dv x Di interaction also indicate that the simple main effect of data set interdependency is significant for both dependent measures when data set size variability is high. Less window management time and fewer window management operations were needed to complete tasks when the interdependency among data sets was low (85.02 s and 14.81 operations) than when it was high (96.88 s and 16.25 operations). The significant simple A x Di interaction, however, indicates that this effect is valid only for the global approach. Performance of the localized approach is insensitive to changes in data set interdependency when data set size variability is high.

The interaction of window management approach with data set size variability is significant for both measures. Figures 11 and 12 depict this interaction for management time and number of management

Table 5. Multiple t-tests on the Simple ANOVA Interaction for High Data Set Size Variability (less than 10 displayed data sets)

a. Window Management Time (sec)

Data Set Interdependency

	Low	High
Approach		
Localized	89.94	90.01
Global	80.19	103.75

p = 0.0478 p = 0.0124

b. Number of Window Management Operations

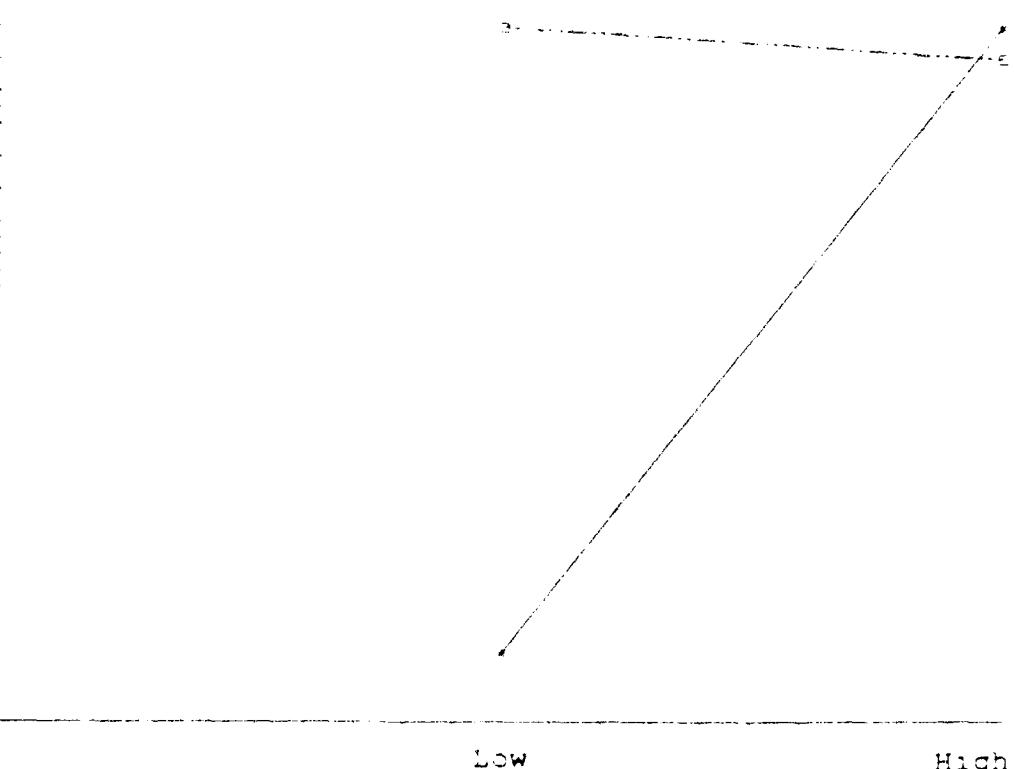
Data Set Interdependency

	Low	High
Approach		
Localized	15.25	15.50
Global	14.38	15.00

p = 0.0216 p = 0.0015

WILLOW MULCH HIGH (H)

Willow mulch was applied at a rate of 1.5 tons per acre. This treatment was applied to the soil surface in the fall of 1987.



DATA SET 1 OF 1 APR 4 9 87
Slope = 0.3381 Intercept = 6.661111

Figure 1. A linear relationship for Willow Management Time (in months) vs. Willow Management.

INITIAL AND WITHDRAWN MACHINING RATES (%)
vs. VARIATION IN THE NUMBER OF CUTTING CYCLES
FOR THE 100% DUTY CYCLE TEST

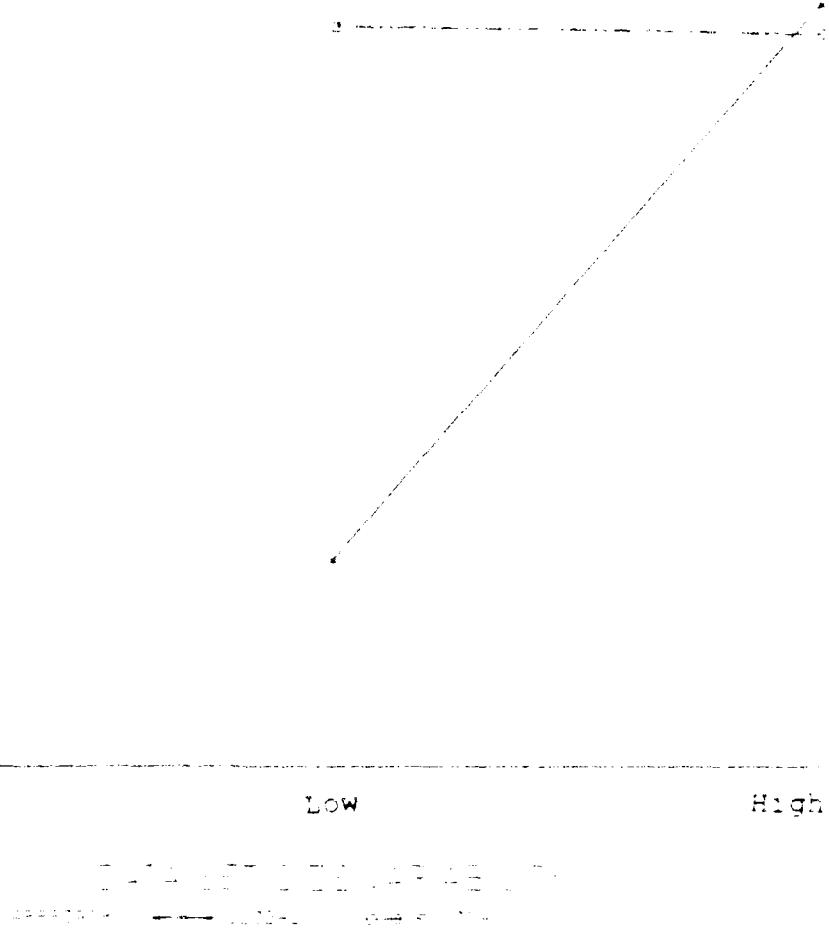


Fig. 10. Effect of variation in the number of cutting cycles on the variation in the initial and withdrawn machining rates for the 100% duty cycle test.

operations, respectively. Multiple t-tests show that the global approach results in less window management time and fewer window management operations than the localized approach when data set size variability is low, but is not significantly different from the localized approach when data set size variability is high (Table VII). The underlying A x Dv x Di interaction indicates, however, that the latter result is confounded by data set interdependency. Specifically, the multiple t tests of Table VI indicate that when data set size variability is high, the global approach results in less window management time and fewer window management operations than the localized approach if data set interdependency is low. When both data set size variability and data set interdependency are high, the global approach results in more management time and operations than the localized approach. The significant A x Dv interaction depicted in Figures 11 and 12 suggests that when there are less than 10 data sets on display, performance with the global approach is more sensitive to changes in data set size variability than performance with the localized approach.

The main effects of all three independent variables are significant for both window management time and number of window management operations. Less window management time and fewer window management operations were needed to complete tasks for the global approach (72.18 s and 11.84 operations) than for the localized approach (90.93 s and 15.41 operations). The underlying A x Dv x Di interaction indicates, however, that this effect does not hold generally. The localized approach is superior to the global approach if data set size variability and data set interdependency are high (see Figures 11 and 12).

Table VII - Multiple t-tests on the A x Inv Interactive for Window Management Times and Number of Window Management Operations (less than 100 displayed Data Sets)

		Data Set Size Variability	
		Low	High
Approach	Localized	91.89	89.98
	Global	50.63	41.93

$p = 0.0001 \quad p = 0.6002$

c) Number of Window Management Operations

		Data Set Size Variability	
		Low	High
Approach	Localized	13.44	13.38
	Global	11.50	13.69

$p = 0.6001 \quad p = 0.5618$

Less window management time and fewer window management operations were needed to complete tasks when the data set size variability was low (11.76 s and 12.32 operations) than when it was high (30.48 s and 12.70 operations). The underlying A x Dv interaction indicates, however, that this effect does not hold generally. It is due to the superior performance of the global approach when data set size variability is low (see Figures 11 and 12).

Less window management time and fewer window management operations were needed to complete tasks when the interdependency among data sets was low (17.00 s and 13.56 operations) than when it was high (86.70 s and 14.69 operations). The underlying A x Dv x Di interaction suggests that this result is due primarily to improvements in the performance of the global approach at high data set size variability when interdependency is low rather than high.

Ten or More Displayed Data Sets

Tables VIII and IX contain the ANOVA summary tables for window management time and number of window management operations, respectively. The interaction of window management approach with data set interdependency is significant for both measures. Figures 13 and 14 depict this interaction for management time and number of management operations, respectively. Multiple t-tests show that the localized approach results in less window management time and fewer window management operations at both levels of task interdependency (Table X). The significant interaction indicates, however, that the advantage of the localized approach over the global approach is greater when the tasks are highly interdependent. When there are ten or more data sets on the display, performance with the global approach appears to be marginally better than with the localized approach.

Table VIII. ANOVA Summary Table for Windows Management Line Length Model
Displayed Data Sets

Source	df	SS	F	p
<u>Between</u>				
Subjects x Sub	1	201923.534		
<u>Within</u>				
Approach (A)	2	296437.54	30.77	0.0003
A x Sub	2	46980.91		
Data Set Size				
Variability (Dv)	2	41.28	0.03	0.8619
Dv x Sub	2	8859.78		
Data Set Inter- dependency (Di)	2	453400.21	28.911	0.0001
Di x Sub	2	12248.75		
A x Dv	1	1907.29	0.53	0.4919
A x Dv x Sub	2	25397.60		
A x Di	1	30422.34	24.42	0.0017
A x Di x Sub	2	8718.79		
Dv x Di	1	99.65	0.37	0.5376
Dv x Di x Sub	2	15404.85		
A x Dv x Di	1	308.65	0.08	0.7810
A x Dv x Di x Sub	2	28391.49		
Total	63			

Table IX ANOVA Summary Table for Number of Window Management Operations (Ten or More Displayed Data Sets)

Source	df	SS	F	P
<u>Between</u>				
Subjects (Sub)	7	439.75		
<u>Within</u>				
Approach (A)	1	2652.25	111.67	0.0001
A x Sub	7	166.25		
Data Set Size				
Variability (Dv)	1	25.00	2.29	0.1142
Dv x Sub	7	76.50		
Data Set Inter-dependency (Di)				
Di	1	17226.56	480.54	0.0001
Di x Sub	7	250.94		
A x Dv	1	6.25	0.38	0.5581
A x Dv x Sub	7	115.75		
A x Di	1	370.56	12.33	0.0098
A x Di x Sub	7	210.44		
Dv x Di	1	0.56	0.04	0.8415
Dv x Di x Sub	7	91.44		
A x Dv x Di	1	0.06	0.00	0.9643
A x Dv x Di x Sub	7	203.44		
Total	63			

WITHIN WINDOW MANAGEMENT TIME (sec.)

DATA SET INDEPENDENCE

APPENDIX ————— 1.0000 0.000000

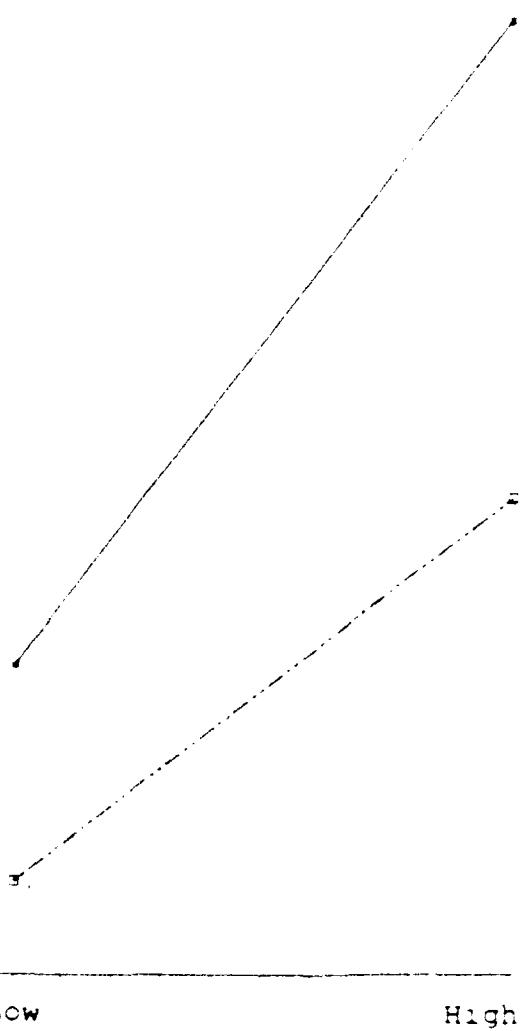
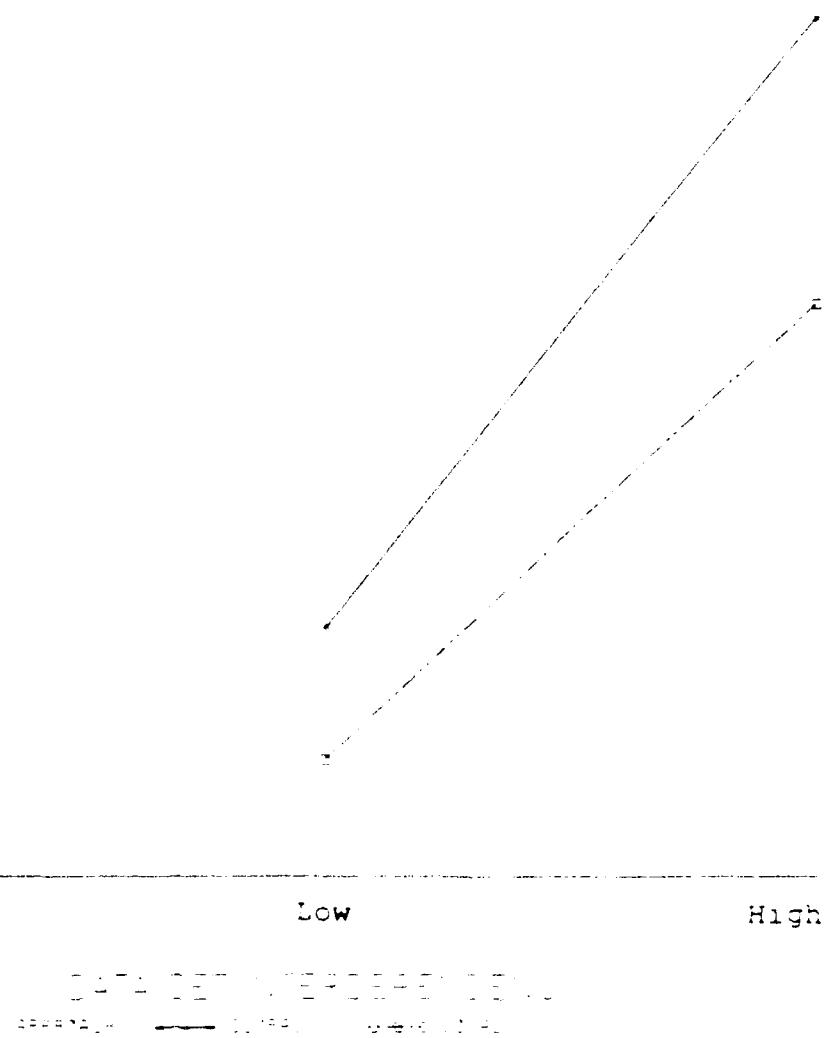


FIGURE 12. A set of data showing the Window Management Time versus Data Set Independence.

FIGURE 4. WILLOW MALLARD DIAKINETIC

Diagram showing the relationship between the time required for the removal of organic material from the water and the concentration of organic material.



Legend: — High — Low

Notes: A small amount of organic material can be removed quickly at low concentrations. At high concentrations, the removal process becomes much slower.

Table X. Multiple t tests on the effect of Interdependent Window Management Times and Number of Window Management Operations for One or More Displayed Data Sets.

Approach Data Set Interdependency
 Low High
 t p

Approach	Data Set Interdependency	
	Low	High
Localized	-2.21 .31	-3.51 .01
Global	-1.77 .31	-3.09 .01
	$p = .0008$	$p = .0001$

b) Number of Window Management Operations

Approach Data Set Interdependency
 Low High
 t p

Approach	Data Set Interdependency	
	Low	High
Localized	-1.46 .60	-2.24 .63
Global	-2.24 .63	-2.21 .71
	$p = .0001$	$p = .0001$

sensitive to changes in data set interdependency than performance with the localized approach.

The main effects of window management approach and data set interdependency are also significant for window management time and number of window management operations. Less window management time and fewer window management operations were needed to complete tasks under the localized approach (289.68 s and 40.63 operations) than under the global approach (403.29 s and 73.50 operations). Less management time and fewer management operations were needed to complete tasks when the interdependency among data sets was low (262.31 s and 50.66 operations) than when it was high (430.65 s and 83.47 operations).

Entire Task Block

Tables XI and XII contain the ANOVA summary tables for window management time and number of window management operations, respectively. The interaction of window management approach with data set interdependency is significant for both dependent measures. Figures 15 and 16 depict this interaction for management time and number of management operations, respectively. Multiple t-tests show that the localized approach results in less window management time and fewer window management operations than the global approach for both low and high data set interdependency tasks (Table XIII). The interaction indicates that the advantage of the localized approach over the global approach is greater when the tasks are highly interdependent. Figures 15 and 16 also suggest that performance with the global approach is more sensitive to changes in data set interdependency than performance with the localized approach.

Table XI - ANOVA Summary Table for Window Management Time Within Task Block.

Source	df	SS	F	p
<u>Between</u>				
Subjects (Sub)	7	303180.48		
<u>Within</u>				
Approach (A)	1	145772.19	17.03	0.0043
A x Sub	7	59223.51		
Data Set Size				
Variability (Dv)	1	4401.16	3.13	0.1203
Dv x Sub	7	9848.26		
Data Set Inter-dependency (Di)	1	507170.09	373.75	0.0001
Di x Sub	7	9498.78		
A x Dv	1	15402.36	3.93	0.0877
A x Dv x Sub	7	27399.51		
A x Di	1	34117.32	26.03	0.0014
A x Di x Sub	7	3174.05		
Dv x Di	1	347.96	0.10	0.7618
Dv x Di x Sub	7	24525.00		
A x Dv x Di	1	3053.53	0.63	0.4525
A x Dv x Di x Sub	7	33786.36		
Total	63			

Table XII ANOVA Summary Table for Number of Window Management Operations (Entire Task Block)

Source	df	SS	F	p
<u>Between</u>				
Subjects (Sub)	1	658.75		
<u>Within</u>				
Approach (A)	1	1701.56	46.38	0.0001
A x Sub	7	190.94		
Data Set Size				
Variability (Dv)	1	39.06	3.79	0.0959
Dv x Sub	7	73.94		
Data Set Inter-dependency (Di)				
Di	1	18428.06	492.47	0.0001
Di x Sub	7	261.94		
A x Dv	1	196.00	14.22	0.0070
A x Dv x Sub	7	96.50		
A x Di	1	400.00	21.79	0.0023
A x Di x Sub	7	128.50		
Dv x Di	1	4.00	0.18	0.6845
Dv x Di x Sub	7	156.00		
A x Dv x Di	1	18.06	0.66	0.4443
A x Dv x Di x Sub	7	192.44		
Total		63		

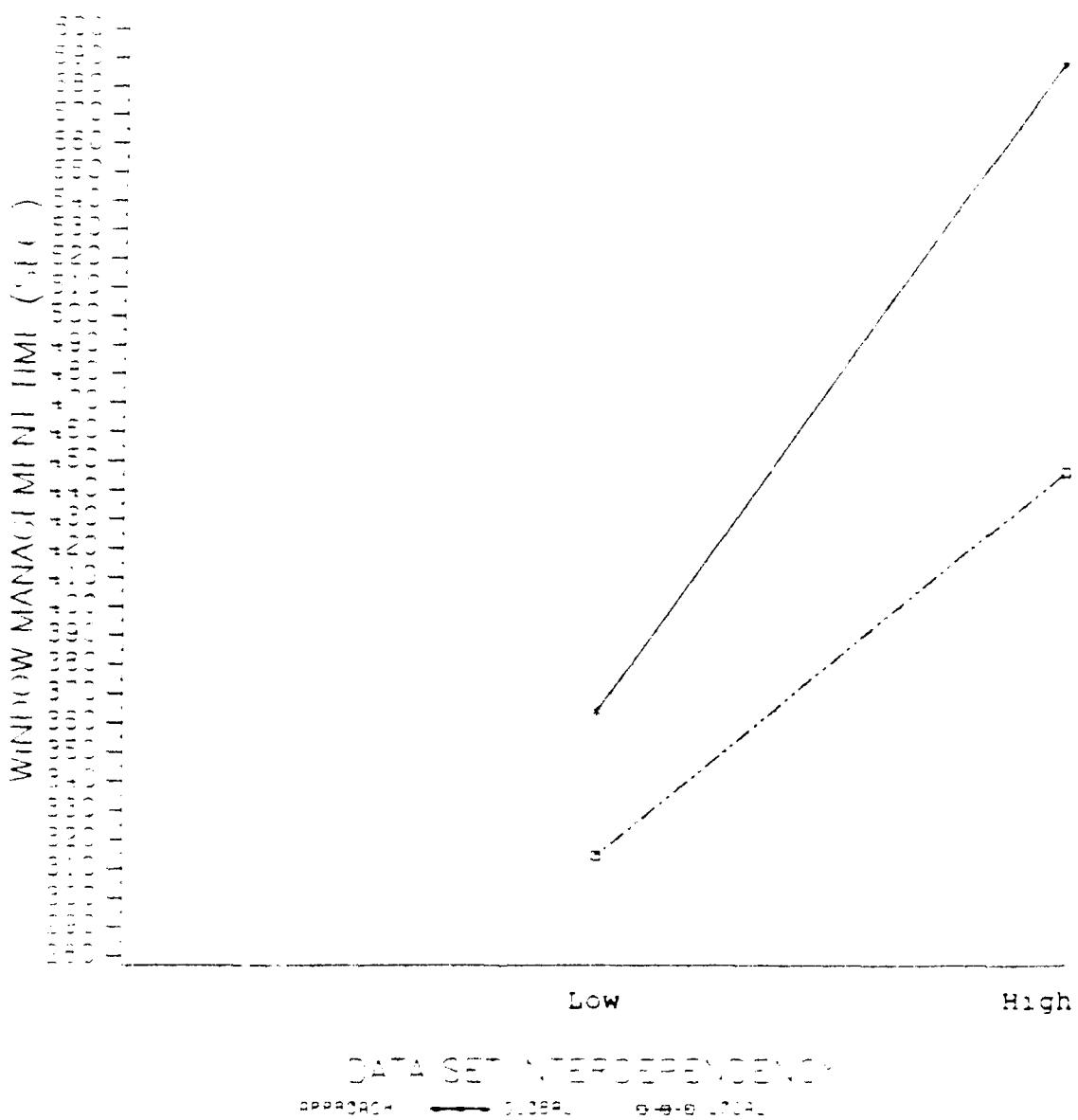


Figure 11. A y-t interaction for Window Management Time - entire Task Block.

NUMBER OF WILLOW MANAGEMENT OPERATIONS
IN THE PAST 12 MONTHS

WILLOW MANAGEMENT OPERATIONS

WILLOW MANAGEMENT OPERATIONS

WILLOW MANAGEMENT OPERATIONS

WILLOW MANAGEMENT OPERATIONS

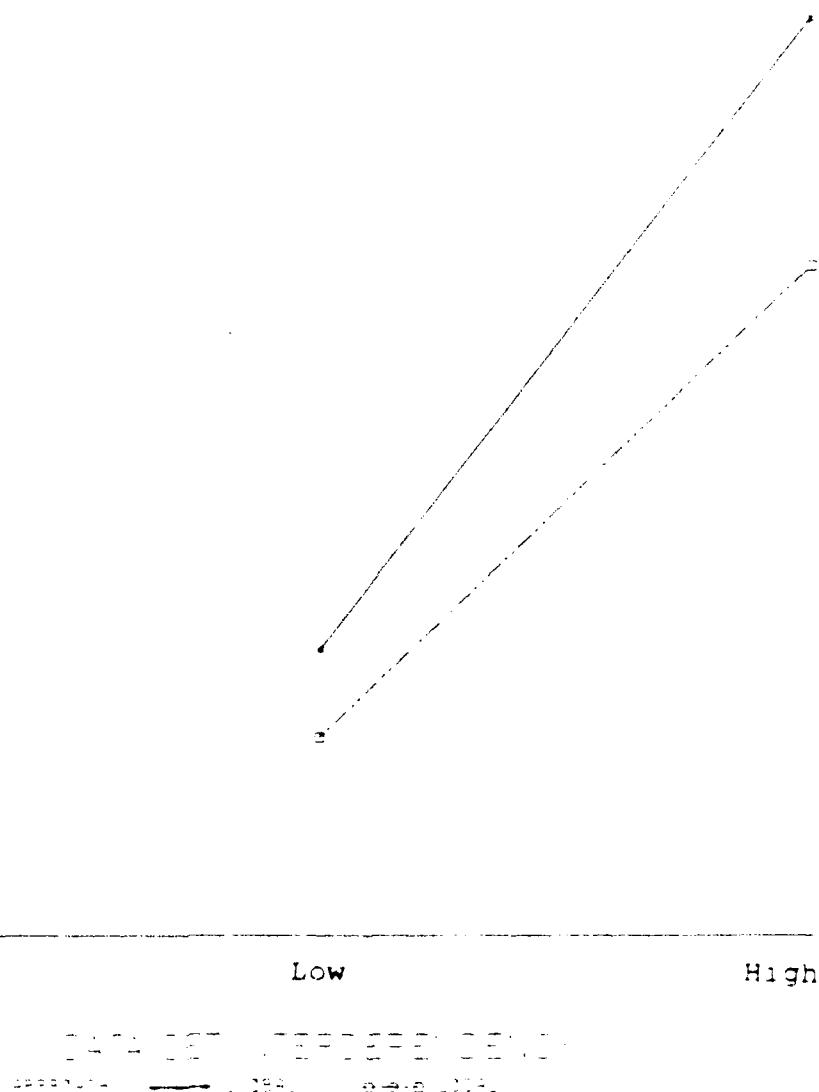


FIGURE 10. A 12-MONTH PERSPECTIVE ON THE NUMBER OF WILLOW MANAGEMENT OPERATIONS IN THE PAST 12 MONTHS.

Figure 10. A schematic diagram of the relationship between the three types of β -converting enzymes and their substrates.

Two types of β -converting enzymes are involved in the conversion of β -amino acids to α -amino acids. One type is the β -amino acid oxidase which converts β -amino acids to α -ketoadic acids. The other type is the β -amino acid reductase which converts β -amino acids to α -amino acids.

Enzymes involved in the conversion of β -amino acids to α -amino acids

β -amino acid oxidase

The β -amino acid oxidase converts β -amino acids to α -ketoadic acids. The reaction is:



(R = H , CH_3 , CH_2CH_3 , etc.)

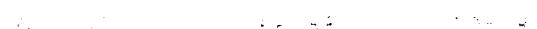
The reaction is catalyzed by O_2 and H_2O .

The β -amino acid oxidase is found in many microorganisms and plants.

Enzymes involved in the conversion of β -amino acids to α -amino acids

β -amino acid reductase

The β -amino acid reductase converts β -amino acids to α -amino acids. The reaction is:



(R = H , CH_3 , CH_2CH_3 , etc.)

The reaction is catalyzed by H_2O_2 and NADH_2 .

The β -amino acid reductase is found in many microorganisms and plants.

The interaction of window management approach and data set size variability is significant for the number of window management operations only. Figure 11 depicts this interaction. Multiple t-tests show that the localized approach results in fewer window management operations than the global approach for both low and high data set size variability tasks (Table XIV). Figure 11 suggests that the localized approach is particularly advantageous when the variability of data set sizes is high. It also suggests that performance with the global approach is more sensitive to changes in data set size variability than performance with the localized approach.

The main effects of window management approach and data set interdependency are also significant for window management time and number of window management operations. Less window management time and fewer window management operations were needed to complete tasks under the localized approach (180.61 s and 76.03 operations) than under the global approach (416.61 s and 86.34 operations). Less management time and fewer management operations were needed to complete tasks when the interdependency among data sets was low (339.32 s and 64.21 operations) than when it was high (311.16 s and 98.16 operations).

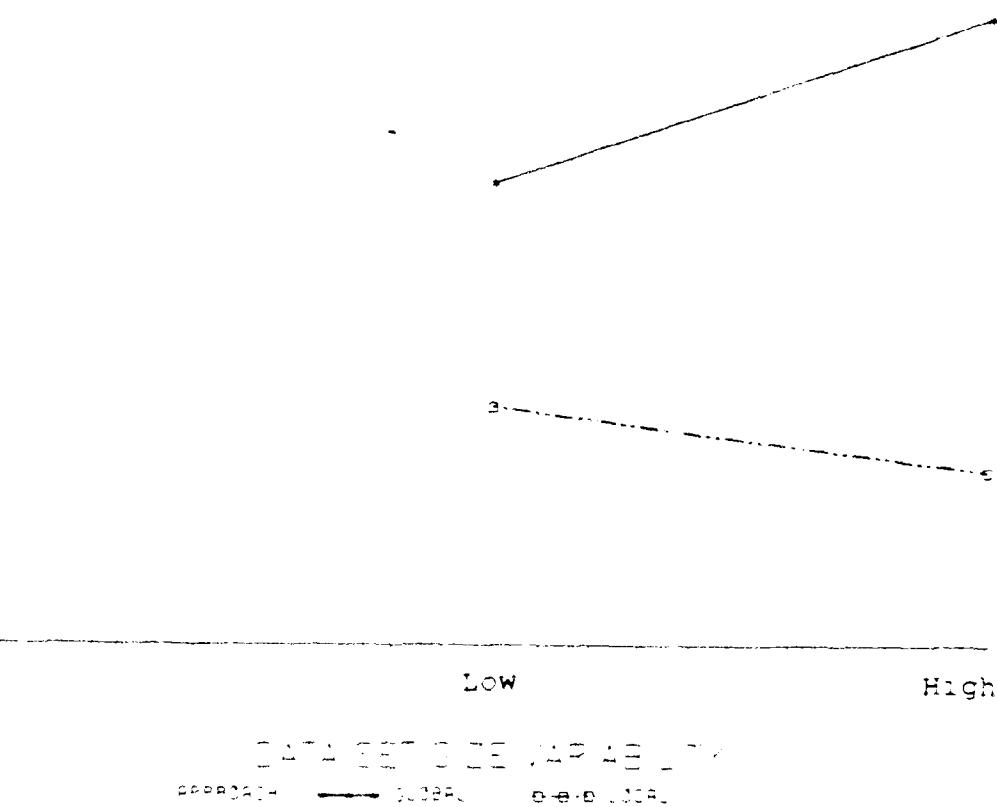
Task Solution Time

Less Than 10 Displayed Data Sets

Table XV contains the ANOVA summary table for the task solution time. The task solution time measures the time subjects devoted to determining the specific tasks requested. It does not include the time devoted to determining a window configuration in which these tasks may be readily contained. The dependent variable and independent variables are the same as before.

NUMBER OF WILLOW MANAGERS ON PASTURE

Number of managers per pasture in each category of number of managers per pasture



Legend: Series 1: Number of Willow Managers on Pasture
Series 2: Number of Managers per Pasture

Table XIV. Multiple t-tests on the A x B interaction for Number of Window Management Operations - Entire Task Block.

Data Set Size Variability		
	Low	High
Localized	77.00	12.05
Application	11.00	12.05
Total	31.81	48.88

$p = .00013$ $p = .0001$

Table IV. ANOVA Summary Table for Task Duration Data Across Three Displayed Data Sets

Source	df	S	F	p
<u>Between</u>				
Subjects x Sub	2	301463.96		
<u>Within</u>				
Approach / A x Sub	1	1.98	6.60	0.9884
606.9 14				
Data Set Size Variability (DV)	2	3846.14	10.56	0.0009
Dv x Sub	2	22361.40		
Data Set Inter- Dependency (Di)	1	62114.06	12.96	0.0087
Di x Sub	2	33613.52		
A x Dv	1	8180.36	2.35	0.1692
A x Dv x Sub	2	24369.04		
A x Di	1	204.63	0.10	0.7636
A x Di x Sub	2	14566.96		
Dv x Di	1	817.10	0.89	0.4010
Dv x Di x Sub	2	7155.21		
A x Dv x Di	1	174.70	0.02	0.8992
A x Dv x Di x Sub	2	70837.59		
Total	63			

The main effect of data set size variability is significant. Less time was needed to solve tasks when the data set size variability was low (415.16 s) than when it was high (515.68 s). The main effect of data set interdependency is also significant. Solution times were shorter when the interdependency among data sets was low (461.30 s) than when it was high (523.66 s). Predictably, window management approach has no effect on the time devoted to determining the requested information on ϵ a suitable window configuration has been achieved.

Ten or More Displayed Data Sets

Table XVI presents the ANOVA summary table for task solution time when ten or more data sets had been displayed. There are no significant interaction effects.

The main effect of data set size variability is significant. Less time was needed to solve tasks when the data set size variability was low (503.28 s) than when it was high (634.28 s). The main effect of data set interdependency is also significant. Solution times were shorter when the interdependency among data sets was low (494.78 s) than when it was high (672.76 s). Again, window management approach has no effect on the time devoted to determining the requested information.

Entire Task Block

Table XVII contains the ANOVA summary table for the entire block of task solution times. There are no significant interaction effects.

The main effect of data set size variability is significant. Less time was needed to solve tasks when the data set size variability was low (415.16 s) than when it was high (515.68 s). The main effect of data set interdependency is also significant. Solution times were

Table XVI. ANOVA Summary of the Data from the Data Set with Multiple Displayed Data Sets

Source	df	SS	F	P
<u>Between</u>				
Subjects x Sub	1	548114.08		
<u>Within</u>				
Approach (A)	1	1873.87	4.19	0.0214
A x Sub	1	141627.81		
Data Set Size	1			
Variability (Dv)	1	163311.36	31.38	0.0013
Dv x Sub	1	45044.19		
Data Set Inter- dependency (Di)	1	506810.51	22.30	0.0022
Di x Sub	1	149103.36		
A x Dv	1	3926.12	0.61	0.4615
A x Dv x Sub	1	45291.82		
A x Di	1	2595.28	0.56	0.5032
A x Di x Sub	1	36482.53		
Dv x Di	1	3834.86	1.08	0.3330
Dv x Di x Sub	1	48825.47		
A x Dv x Di	1	1.03	0.06	0.9949
A x Dv x Di x Sub	1	168145.59		
Total	3			

Table AIII - ANOVA Summary Table for Case Selection Time Analysis
Block

Source	df	SS	F	p
<u>Between</u>				
Subjects x sub	1	16135.493		
<u>Within</u>				
Approach A	1	1712.31	1.07	0.304
A x Sub	1	3712.0848		
Data Set Size	1			
Variability x Dv	1	492160.13	49.81	0.00014
Dv x Sub	1	115401.24		
Data Set Inter- dependency (Di)	1	924201.03	10.50	0.0017
Di x Sub	1	315592.51		
A x Dv	1	26440.76	1.35	0.2836
A x Dv x Sub	1	121664.80		
A x Di	1	4258.05	0.23	0.6268
A x Di x Sub	1	84292.44		
Dv x Di	1	8192.29	0.18	0.3136
Dv x Di x Sub	1	48651.62		
A x Dv x Di	1	148.87	0.00	0.9817
A x Dv x Di x Sub	1	442672.39		
Total	63			

shortened when the interaction of window data set size was low, but not when it was high (Table 18). Window management approach has no effect on the time devoted to determining the requested information.

Number of Incorrect Solutions

Tables XVIII, XIX, and XX present the ANOVA summary tables for the number of incorrect task solutions when less than ten data sets had been displayed in the screen, for the remainder of the task block, and for the entire task block, respectively. None of the independent variables, either singly or in combination, affected the number of incorrect task solutions. Table XVI presents the mean numbers of incorrect task solutions entered in each treatment condition for the initial portion of the task block, the remainder of the task block, and the entire task block. The mean numbers of incorrect solutions are generally low, not exceeding two per task block in any of the treatment conditions (out of 36 total solutions per task block). The maximum number of incorrect solutions entered by a subject in a treatment condition was four.

Total Task Completion Time

Less Than 10 Displayed Data Sets

Table XXI contains the ANOVA summary table for the total time required to complete the tasks. This measure is the sum of the window management time and the task solution time. The interaction of window management approach with data set size variability is significant (Table 19) depicts this interaction. Multiple t-tests show that less time was required to complete the tasks with the global approach than with the localized approach when data set size variability was low, but that task completion times were not significantly different for the two

Table F(11). ANOVA Summary Table for Number of Different Solutions Less Than 10 Displayed Data Sets

Source	df	SS	F	p
<u>Between</u>				
Subjects x Sub.	1	5.71		
<u>Within</u>				
Approach A	1	0.00	0.24	0.7226
A x Sub	1	0.14		
Data Set Size	1			
Variability DV	1	0.00	0.00	1.0000
Dv x Sub	1	0.25		
Data Set Inter- dependency - Di	1	0.20	0.70	0.4304
Di x Sub	1	2.50		
A x Dv	1	0.06	0.08	0.7895
A x Dv x Sub	1	5.63		
A x Di	1	0.06	0.37	0.5630
A x Di x Sub	1	1.19		
Dv x Di	1	0.00	0.30	1.0000
Dv x Di x Sub	1	5.25		
A x Dv x Di	1	0.06	0.26	0.6263
A x Dv x Di x Sub	1	1.69		
Total	63			

Table XIX. ANOVA Summary Table for Number of Incorrect Solutions (Ten or More Displayed Data Sets)

Source	df	SS	F	P
<u>Between</u>				
Subjects (Sub)	7	28.37		
<u>Within</u>				
Approach (A)	1	0.14	0.14	0.6776
A x Sub	7	5.23		
Data Set Size				
Variability (Dv)	1	0.92	0.04	0.8505
Dv x Sub	7	2.86		
Data Set Inter-dependency (Di)				
Di	1	0.39	0.26	0.6253
Di x Sub	7	10.48		
A x Dv	1	0.02	0.05	0.8357
A x Dv x Sub	7	2.36		
A x Di	1	0.02	0.03	0.8786
A x Di x Sub	7	4.36		
Dv x Di	1	0.02	0.02	0.8850
Dv x Di x Sub	7	4.86		
A x Dv x Di	1	0.02	0.05	0.8357
A x Dv x Di x Sub	7	2.36		
Total	63			

Table XX ANOVA Summary Table for Number of Incorrect Solutions - Entire Task Block

Source	df	SS	F	p
<u>Between</u>				
Subjects (Sub)	7	45.6		
<u>Within</u>				
Approach (A)	1	0.02	0.02	0.9013
A x Sub	7	5.12		
Data Set Size				
Variability (Dv)	1	0.02	0.02	0.8878
Dv x Sub	7	5.12		
Data Set Interdependency (Di)				
Di x Sub	1	0.02	0.01	0.9298
Di x Sub	7	13.12		
A x Dv	1	0.02	0.02	0.9013
A x Dv x Sub	7	6.61		
A x Di	1	0.14	0.28	0.6115
A x Di x Sub	7	3.48		
Dv x Di	1	0.02	0.01	0.9335
Dv x Di x Sub	7	14.61		
A x Dv x Di	1	0.14	0.02	0.6702
A x Dv x Di x Sub	7	4.98		
Total		63		

Table XXI Mean Numbers of Incorrect Solutions

a) Less than 10 displayed data sets

	Approach			
	Localized	Global	Localized	Global
	Low Inter-dependency	High Inter-dependency	Low Inter-dependency	High Inter-dependency
Low Size Variability	0.75	0.75	0.88	0.63
High Size Variability	0.75	0.63	0.88	0.75

b) Ten or more displayed data sets

	Approach			
	Localized	Global	Localized	Global
	Low Inter-dependency	High Inter-dependency	Low Inter-dependency	High Inter-dependency
Low Size Variability	1.13	1.38	1.13	1.25
High Size Variability	1.25	1.38	1.13	1.25

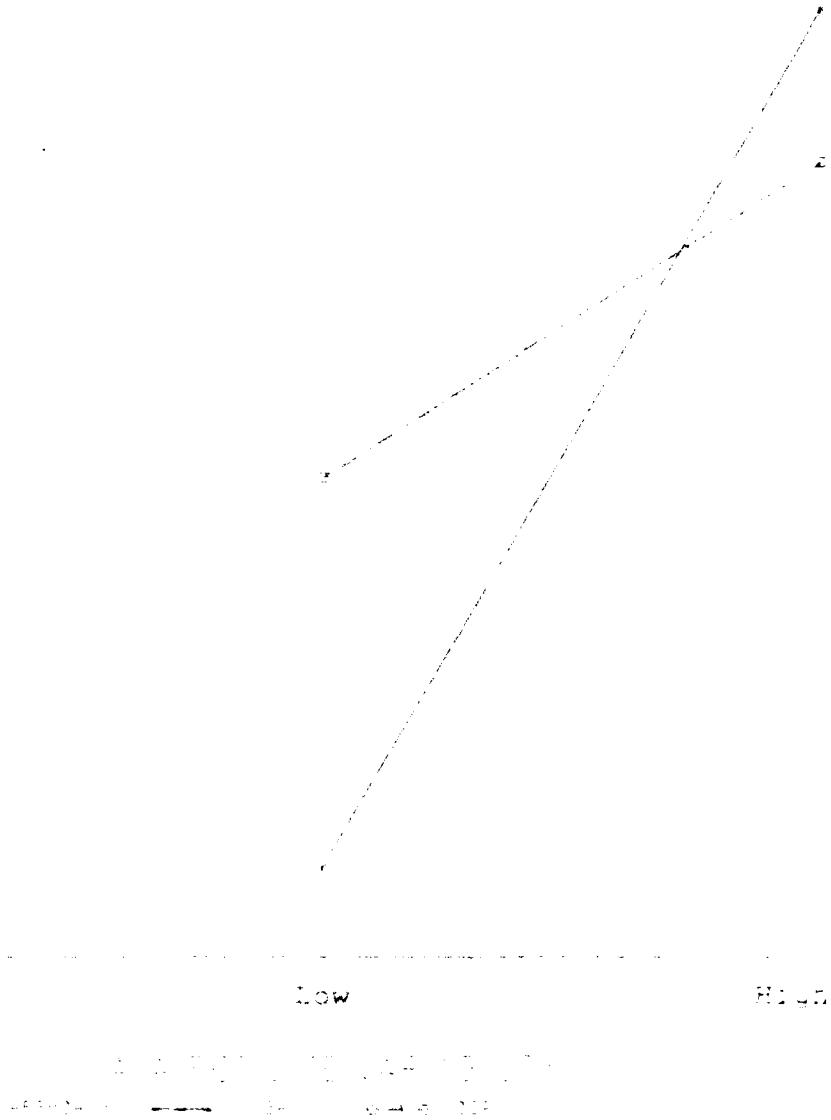
c) Entire task block

	Approach			
	Localized	Global	Localized	Global
	Low Inter-dependency	High Inter-dependency	Low Inter-dependency	High Inter-dependency
Low Size Variability	1.93	2.10	2.06	1.86
High Size Variability	2.36	2.70	2.50	2.36

Table XXII. ANOVA Summary Table for Task completion time less than
Displayed Data Sets

Source	df	SS	F	p
<u>Between</u>				
Subjects x Sub	1	340703.57		
<u>Within</u>				
Approach (A)	1	5480.81	1.51	0.496
A x Sub	1	14682.86		
Data Set Size				
Variability (Dv)	1	137048.04	47.68	0.0002
Dv x Sub	1	20118.74		
Data Set Interdependency (Di)				
Di x Sub	1	83083.74	17.26	0.0043
A x Dv	1	29194.85	7.03	0.0329
A x Dv x Sub	1	29072.53		
A x Di	1	16.02	0.01	0.9320
A x Di x Sub	1	14318.38		
Dv x Di	1	1387.56	0.12	0.4254
Dv x Di x Sub	1	13562.96		
A x Dv x Di	1	559.09	0.05	0.8303
A x Dv x Di x Sub	1	79056.64		
Total	63			

FIGURE 1. (Left) Effect of Na^+ concentration on the rate of polymerization of styrene at 50°C . The reaction was carried out in benzene solution containing 1 mole/liter of FeCl_3 , 1 mole/liter of H_2O_2 , and 1 mole/liter of NaCl . The polymerization time was 1 hour.



(Right) Effect of Na^+ concentration on the rate of polymerization of styrene at 50°C . The reaction was carried out in benzene solution containing 1 mole/liter of FeCl_3 , 1 mole/liter of H_2O_2 , and 1 mole/liter of NaCl . The polymerization time was 1 hour.

spur values when the display size variability was high. Table 4(1) shows the effect of the independent variables on the total time taken to complete the tasks. The display performance with the global approach was more efficient than changes in display size, variability than performance with the local approach.

Table 4(1). Analysis of variance for the total time taken to complete the tasks when less than 10% display data was lost

		Aptitude			
		Individual		Group	
		Mean	S.E.M.	Mean	S.E.M.
Display	Global	10.84	0.12	10.14	0.09
Task	1	10.84	0.12	10.14	0.09
Task	2	10.84	0.12	10.14	0.09
Task	3	10.84	0.12	10.14	0.09
Task	4	10.84	0.12	10.14	0.09
Task	5	10.84	0.12	10.14	0.09
Task	6	10.84	0.12	10.14	0.09
Task	7	10.84	0.12	10.14	0.09
Task	8	10.84	0.12	10.14	0.09
Task	9	10.84	0.12	10.14	0.09
Task	10	10.84	0.12	10.14	0.09
Task	11	10.84	0.12	10.14	0.09
Task	12	10.84	0.12	10.14	0.09
Task	13	10.84	0.12	10.14	0.09
Task	14	10.84	0.12	10.14	0.09
Task	15	10.84	0.12	10.14	0.09
Task	16	10.84	0.12	10.14	0.09
Task	17	10.84	0.12	10.14	0.09
Task	18	10.84	0.12	10.14	0.09
Task	19	10.84	0.12	10.14	0.09
Task	20	10.84	0.12	10.14	0.09
Task	21	10.84	0.12	10.14	0.09
Task	22	10.84	0.12	10.14	0.09
Task	23	10.84	0.12	10.14	0.09
Task	24	10.84	0.12	10.14	0.09
Task	25	10.84	0.12	10.14	0.09
Task	26	10.84	0.12	10.14	0.09
Task	27	10.84	0.12	10.14	0.09
Task	28	10.84	0.12	10.14	0.09
Task	29	10.84	0.12	10.14	0.09
Task	30	10.84	0.12	10.14	0.09
Task	31	10.84	0.12	10.14	0.09
Task	32	10.84	0.12	10.14	0.09
Task	33	10.84	0.12	10.14	0.09
Task	34	10.84	0.12	10.14	0.09
Task	35	10.84	0.12	10.14	0.09
Task	36	10.84	0.12	10.14	0.09
Task	37	10.84	0.12	10.14	0.09
Task	38	10.84	0.12	10.14	0.09
Task	39	10.84	0.12	10.14	0.09
Task	40	10.84	0.12	10.14	0.09
Task	41	10.84	0.12	10.14	0.09
Task	42	10.84	0.12	10.14	0.09
Task	43	10.84	0.12	10.14	0.09
Task	44	10.84	0.12	10.14	0.09
Task	45	10.84	0.12	10.14	0.09
Task	46	10.84	0.12	10.14	0.09
Task	47	10.84	0.12	10.14	0.09
Task	48	10.84	0.12	10.14	0.09
Task	49	10.84	0.12	10.14	0.09
Task	50	10.84	0.12	10.14	0.09
Task	51	10.84	0.12	10.14	0.09
Task	52	10.84	0.12	10.14	0.09
Task	53	10.84	0.12	10.14	0.09
Task	54	10.84	0.12	10.14	0.09
Task	55	10.84	0.12	10.14	0.09
Task	56	10.84	0.12	10.14	0.09
Task	57	10.84	0.12	10.14	0.09
Task	58	10.84	0.12	10.14	0.09
Task	59	10.84	0.12	10.14	0.09
Task	60	10.84	0.12	10.14	0.09
Task	61	10.84	0.12	10.14	0.09
Task	62	10.84	0.12	10.14	0.09
Task	63	10.84	0.12	10.14	0.09
Task	64	10.84	0.12	10.14	0.09
Task	65	10.84	0.12	10.14	0.09
Task	66	10.84	0.12	10.14	0.09
Task	67	10.84	0.12	10.14	0.09
Task	68	10.84	0.12	10.14	0.09
Task	69	10.84	0.12	10.14	0.09
Task	70	10.84	0.12	10.14	0.09
Task	71	10.84	0.12	10.14	0.09
Task	72	10.84	0.12	10.14	0.09
Task	73	10.84	0.12	10.14	0.09
Task	74	10.84	0.12	10.14	0.09
Task	75	10.84	0.12	10.14	0.09
Task	76	10.84	0.12	10.14	0.09
Task	77	10.84	0.12	10.14	0.09
Task	78	10.84	0.12	10.14	0.09
Task	79	10.84	0.12	10.14	0.09
Task	80	10.84	0.12	10.14	0.09
Task	81	10.84	0.12	10.14	0.09
Task	82	10.84	0.12	10.14	0.09
Task	83	10.84	0.12	10.14	0.09
Task	84	10.84	0.12	10.14	0.09
Task	85	10.84	0.12	10.14	0.09
Task	86	10.84	0.12	10.14	0.09
Task	87	10.84	0.12	10.14	0.09
Task	88	10.84	0.12	10.14	0.09
Task	89	10.84	0.12	10.14	0.09
Task	90	10.84	0.12	10.14	0.09
Task	91	10.84	0.12	10.14	0.09
Task	92	10.84	0.12	10.14	0.09
Task	93	10.84	0.12	10.14	0.09
Task	94	10.84	0.12	10.14	0.09
Task	95	10.84	0.12	10.14	0.09
Task	96	10.84	0.12	10.14	0.09
Task	97	10.84	0.12	10.14	0.09
Task	98	10.84	0.12	10.14	0.09
Task	99	10.84	0.12	10.14	0.09
Task	100	10.84	0.12	10.14	0.09

The main effect of data set size variability and data set interdependent on the total time required to complete the tasks when the global and local approaches were compared. The completion time when the data set size variability was low (0.8 Gb/s) was higher than when it was high (1.6 Gb/s). The time required to complete the tasks when the data set interdependent was higher than the completion time when it was independent. The completion time was higher when the data set was small (10%) than when it was large (20%).

Indirect Mode Displayed Data Sets

Table XVII contains the ANOVA summary table for the total time required to complete the tasks when ten or more data sets had been displayed on the screen. There are no significant interaction effects.

The main effects of data set size variability and data set interdependence are significant. Less time was needed to complete tasks when the data set size variability was low (817.41 s) than when it was high (971.96 s). Less time was needed to complete tasks when the interdependence among data sets was low (760.32 s) than when it was high (811.18 s).

The main effect of window management approach is nearly significant ($F = 3.0580$). This result suggests that when there are ten or more data sets on the display, less time is needed to complete tasks under the parallel approach (868.10 s) than under the global approach (907.41 s).

Entire Task Block

Table XV contains the ANOVA summary table for the entire block of task completion times. There are no significant interaction effects. The effects of data set size variability and data set interdependence are significant. Less time was needed to complete tasks when the data set size variability was low (1405.18 s) than when it was high (1577.48 s). Less time was needed to complete tasks when the interdependence among data sets was low (1198.91 s) than when it was high (1371.17 s).

Subjective Data

A subjective rating scale questionnaire was completed by each participant at the end of each treatment condition and at the conclusion of

Table XIV. ANOVA Summary Table for Task Completion Time (Sec.) by Mode
Displayed Data Sets

Source	df	SS	F	p
<u>Between</u>				
Subjects Subj	2	144252.75		
<u>Within</u>				
Approach A)	1	247251.35	5.13	0.0280
A X Sub	1	337704.56		
Data Set Size				
Variability Dv	1	128672.12	15.84	0.0053
Dv X Sub	1	78957.19		
Data Set Inter-				
dependency Di	1	1850291.07	93.88	0.0001
Di X Sub	1	137959.27		
A X Dv	1	17247.99	0.78	0.4063
A X Dv X Sub	1	134707.88		
A X Di	1	9636.39	0.53	0.3260
A X Di X Sub	1	75954.81		
Dv X Di	1	5170.87	0.74	0.4185
Dv X Di X Sub	1	48999.71		
A X Dv X Di	1	377.09	0.61	0.4228
A X Dv X Di X Sub	1	261888.44		
Total	63			

Table XXV. ANOVA Summary Table for Park Temperature Change Between and Within Block

Source	df	SS	F	P
<u>Between</u>				
Subjects (Sub)	7	1971333.41		
<u>Within</u>				
Approach (A)	1	178051.30	1.18	0.2695
A X Sub	7	698786.44		
Data Set Size	1	626703.79	26.09	0.0014
Variability (Dv)	7	168161.44		
Dv X Sub	7			
Data Set Inter-dependency (Di)	1	2713428.44	61.82	0.0001
Di X Sub	7	288582.50		
A X Dv	1	92079.63	2.19	0.1742
A X Dv X Sub	7	281872.32		
A X Di	1	9162.08	2.11	0.4972
A X Di X Sub	7	125145.14		
Dv X Di	1	12190.09	0.87	0.3830
Dv X Di X Sub	7	98499.94		
A X Dv X Di	1	1748.39	0.72	0.3916
A X Dv X Di X Sub	7	612844.03		
Total	63			

the study to permit comparison of the two window management approaches on dimensions of perceived ease of learning and satisfaction. Appendix C and D. This questionnaire data was analyzed using the Kruskal-Wallis test (Gibbons, 1985). None of the comparisons of the two window management approaches were found to be significant for any of the treatment conditions.

At the conclusion of the study each participant was also asked to rank the two window management approaches in order of overall preference.

Appendix E Six subjects indicated a preference for the localized approach, while two subjects preferred the global approach. The Friedman test (Gibbons, 1985) was used to analyze this data with window management approach the factor of interest and subjects the blocking factor. The results of this analysis indicate that there is no significant difference in preference between the two window management approaches ($p > 0.1$).

CHAPTER VI

DISCUSSION

The objective of this research was to determine the relative efficiencies of the localized and global approaches to tiled window management under each of several decision making scenarios. The results of the study of human performance using these window management approaches show that their relative efficiencies depend on whether or not the window's display area is large enough, on average, to permit all open windows to display simultaneously the relevant contents of the data sets associated with them. When there is more than enough display area to permit every open window to display the entire contents of its associated data set, the relative efficiencies of the two window management approaches also depend on the variability in the sizes of the data sets required to perform the tasks and on the level of interdependency among these data sets. The results of the study that lead to these general conclusions are discussed in this chapter.

Less Than 10 Displayed Data Sets

In the study reported in Chapters IV and V, a maximum of two rows of windows were permitted on the display. The sizes of the data sets contained within the windows were such that, on average, five data sets could be displayed simultaneously within each of these rows. Thus, on average, the total display area was large enough to permit ten open windows to display simultaneously the entire contents of the data sets associated with them.

The first hypothesis tested in Chapter 10 predicted that when there were less than 10 windows on the display, the global window management approach would be more efficient than the localized approach if the data set size variability was low. The results of the study support this hypothesis. When less than 10 windows were on the display and data set size variability was low, the global approach required less window management time (Figure 9), fewer window management operations (Figure 11), and shorter total task completion times (Figure 18), than the localized approach. This was true whether data set interdependency was low or high.

The basis for the first hypothesis was that when there is enough display area, on average, to permit all open windows to display simultaneously the entire contents of their associated data sets and the data sets are approximately equal in size, it is likely that each of the windows contains space in excess of that required to display the associated data set. Thus, if an existing window needs to be enlarged or a new window needs to be opened, the small amount of space that will be taken from each window to accomplish this operation under the global approach will often leave these windows sufficiently large to continue display of the full associated data sets. Under the localized approach, however, space will be taken from only one or two windows. The large amount of space that each of these windows must give up may leave these windows too small to continue display of the full associated data set. Thus, enlarging an existing window or opening a new window may have to be followed by additional window enlarging operations when the localized window management approach is used.

The study design included a global and a localized window management approach. A maximum of $M = 1$ rows of windows was permitted on the display and an average of $N = 4$ data sets could be completely displayed within each of these rows. Given the basis of the research hypothesis just stated, however, the advantage of the global approach over the localized approach would be expected to hold, and even increase, when N , the average number of data sets that can be completely displayed in each row, is greater than five and there are less than $M \times N$ windows on the display.

The second research hypothesis stated in Chapter III predicted that when there were less than 10 windows on the display, the localized window management approach would be more efficient than the global approach if both the data set size variability and the data set interdependency were high. The results of the study support this hypothesis. When less than 10 windows were on the display and both data set size variability and data set interdependency were high, the localized approach required less window management time (Figure 9) and fewer window management operations (Figure 10) than the global approach.

The basis for the second hypothesis was that when there is enough display area, on average, to permit all open windows to display simultaneously the entire contents of their associated data sets and the data sets vary greatly in size, it is likely that some windows contain a great deal more space than that required to display their associated data sets while other windows do not contain much excess space. If the requirements required to perform a task are highly interdependent, the full contents of several of these data sets may need to be displayed simultaneously to permit performance of a task. If an existing window needs to be enlarged or a new window needs to be opened, the localized

approach permits the user to take space from only one or two windows to meet the needs of the operation and it may be possible to take space only from those windows with sufficient excess space to meet the needs of the operation. Additional enlarging operations may be necessary only if the data sets associated with the affected windows are needed to complete later tasks. Under the global approach, more windows would be affected by the operations, increasing the likelihood that windows without much excess space would be reduced below the size needed to view their associated data sets. This, in turn, would increase the likelihood that additional enlarging operations would be required to complete the current or later tasks when the global management approach is used.

While the study tested a situation in which an average of $N = 5$ data sets could be completely displayed within each row of windows, the global approach would affect even larger numbers of windows when an existing window needs to be enlarged or a new window needs to be opened and the value of N is greater than five. Thus, the advantage of the localized approach would be expected to hold, and even increase, when N is greater than five and there are less than $M \times N$ windows on the display.

No hypotheses were stated for the case in which there are less than $M \times N$ windows on the display, data set size variability is high and data set interdependency is low. The results of the study indicate, however, that when there were less than 10 windows on the display, the global window management approach required less window management time (Figure 7) and fewer window management operations (Figure 10) than the localized approach when data set size variability was high and data set interdependency was low. The difference in performance between the two window management approaches, while significant, was smaller than the

differences observed in the cases for which the research hypotheses provided predictions of the results. Because the basis of the global approach's advantage over the localized approach in this case is not clear and the observed differences are relatively small there is little support for a conjecture that the advantage of the global approach might hold for values of N other than the tested value of five.

Ten or More Displayed Data Sets

When more than ten windows were displayed simultaneously, the total display area was, on average, not large enough to permit viewing of the entire contents of the associated data sets. The third research hypothesis stated in Chapter III predicted that when there were ten or more windows on the display the localized window management approach would be more efficient than the global approach if the interdependency among data sets was high. The results of the study support this hypothesis and suggest that the localized approach is more efficient than the global approach even when the interdependency among data sets is low. When ten or more windows were on the display, the localized approach required less window management time (Figure 13) and fewer window management operations (Figure 14) than the global approach. The advantage of the localized approach over the global approach was significant for both levels of data set interdependency, but was greater when the interdependency among data sets was high. The advantage of the localized approach was not affected by the level of data set size variability.

The basis for the third hypothesis was that when there is not enough display area, on average, to permit all open windows to display simultaneously the entire contents of their associated data sets, it is likely that the subtraction of any space at all from a window will

reduce its size below that required to display the full contents of the associated data set. If an existing window needs to be enlarged or a new window needs to be opened, the localized approach may reduce only one or two windows below the size required to view their full contents. Additional enlarging operations will be necessary only if the data sets contained in these windows are necessary to complete the current task or subsequent tasks. Under the global approach, however, several windows will be reduced below the size needed to view their full contents, increasing the likelihood that additional enlarging operations will be required to complete the current or subsequent tasks. The likelihood that additional enlarging operations will be required under the global approach is particularly high when the data sets required to complete tasks are highly interdependent, because in this case the full contents of several data sets may have to be displayed simultaneously to permit performance of the task.

The study tested a situation in which an average of $N = 5$ data sets could be completely displayed within each row of windows. The global approach would affect even larger numbers of windows when an existing window needs to be enlarged or a new window needs to be opened and the value of N is greater than five. Thus, the advantage of the localized approach is expected to hold, and even increase, when N is greater than five and there are $M \times N$ or more windows on the display.

CHAPTER VII

CONCLUSION

Recommendations

The objective of this research was to determine the relative efficiencies of the localized and global approaches to tiled window management under each of several decision making scenarios. Each of these decision making scenarios is characterized by its location along each of three dimensions:

1. Whether or not the windowed display area is large enough, on average, to permit all open windows to display simultaneously the relevant contents of the data sets associated with them.
2. The variability in the sizes of the data sets required to perform the decision making tasks.
3. The level of interdependency among the data sets required to perform the decision making tasks.

The results of the study of human performance using the localized and global window management approaches led to the following empirically supported guidelines for the design of window management systems:

When there is enough display area, on average, to permit all open windows to display simultaneously the relevant contents of the data sets associated with these windows the following alternatives are recommended:

If the variability in the sizes of the data sets required to perform the decision making tasks is low, the global window management approach is recommended. This recommendation holds both when the interdependency among the data sets required to perform the tasks is low and when it is high.

If both the size variability and the interdependency among the data sets required to perform the decision making tasks are high, the localized window management approach is recommended.

This research does not provide sufficient support for a general recommendation applicable to the case in which the data set size variability is high and the interdependency among data sets is low. For the particular decision making environment tested, however, the global window management approach proved to be superior to the localized approach in this case.

When there is not enough display area, on average, to permit all open windows to display simultaneously the relevant contents of the data sets associated with these windows, the localized window management approach is recommended. This recommendation holds regardless of the levels of size variability and interdependency among the data sets required to perform the decision making tasks.

Limitations and Suggestions for Future Research

The research hypotheses stated in Chapter III were supported empirically by the results of the study reported in Chapters IV and V. The hypotheses are stated for a situation in which a maximum of M rows of windows is permitted on a display and an average of N data sets can be completely displayed within each of these rows. The study reported in Chapters IV and V tested a situation in which M was set equal to 2 and N was set equal to 5. Thus, empirical support for the research hypotheses is provided by the study only for these specific values of M and N. The bases for the research hypotheses suggest, however, that they should hold for values of M other than two and for values of N

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LOCALIZED AND GLOBAL APPROACHES FOR TILED
WINDOW MANAGEMENT BY: JF CHEN, JS GREENSTEIN

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and the results of the research will be discussed below. The final section will discuss the implications of the research for decision making.

The research explicitly considered two-dimensional data sets. This study and previous studies have demonstrated that the performance of decision makers in one-dimensional situations is often improved when the data set is partitioned into two dimensions. For example, when the data set was divided into two half the display height, the search times for windows management system users permit the window manager to view more windows at once. It may also permit the potential user to switch between windows more easily. It may also be adjusted independently of each other. It is possible that the approaches stated in this research could be generalized to the more general situation in which tiled window dimensions may be adjusted in two dimensions rather than in one. It is suggested that future research pursue this possibility.

This research explicitly considered a situation in which the performance of a decision making task required access to the entire contents of a data set. That is, the task required the decision maker to view all the graphic objects in a data set if he was to be certain that he had located all the objects relevant to a particular task. In many realistic decision making situations, only portions of data sets need to be accessed to perform decision making tasks. In such situations it would probably be appropriate to include a window management system. The window management system would permit the user to examine the relevant portion of a data set within a window that does not contain the full data set. Such a large window would permit the presentation of a portion of the data set. Such a portion of the data set would be

in a high number of windows, it would be more difficult to identify which window in a row of windows it would be more appropriate to move. In addition, the storage number of the last opened window would have to be stored in the saved complex entry within a row. It is suggested that future research be conducted to determine whether the hypotheses stated in this research might also apply to this more general situation.

This research assumed a situation in which windows were closed when the enlarging or opening of one window reduced the amount of space available to another window in the same row below a specified minimum acceptable width. Also, when a window was opened, the amount of space allocated to the new window was set equal to $1/M$ of the total row width, where M represents the number of windows present in the row upon introduction of the new window. These design decisions permitted the user to manage windows using a very small set of commands. The user would certainly gain additional flexibility if additional window management functions were introduced, such as a "close window" function, a "replace window" function, and an "open window" function which also permitted the user to specify the width of the new window. This flexibility might increase efficiency by permitting the user to exert greater control over the manner in which space is made available for enlarged and newly opened windows. It is suggested that future research be conducted to determine whether the additional control offered by this additional functionality outweighs the complexity added to the window management task.

The literature review presented in Chapter II identified seven categories of human factors issues directly relevant to the design of auxiliary electronic data computer interfaces. These categories included

example, a task requiring a large amount of information to be processed may require a different technique. This research has demonstrated that these different techniques can be used effectively in the design of window management systems. In addition, it has also shown that different window managers can be evaluated and compared within several different categories and characteristics. Thus, it seems appropriate to consider the following simplified window management system:

Finally, the empirical research of human performance with window management systems has thus far been conducted. Clearly, much research remains to be done before system designers can be provided with a comprehensive set of guidelines for the design and use of window management systems that enhance user task performance and satisfaction.

APPENDICES

INSTITUTE FOR COMPUTER WORKSTATIONS
UNIVERSITY OF CALIFORNIA, BERKELEY

The following document is intended to provide you with information about the experiment and to inform you of your rights as a participant.

This study investigates two different approaches commonly used in window systems. The first, termed the adjacent approach, reserves space for a new or enlarged window by taking space from adjacent windows. When the adjacent windows do not have enough space to meet the needs of the new or enlarged window, space is also taken from the next row of adjacent windows. The second, termed the global approach, creates space for a new or enlarged window by taking equal amounts of space from all the other windows in the same row. The information this study will produce is needed as window management systems are to be employed effectively. This research is being conducted in the Human-Computer Systems Laboratory of the Department of Industrial Engineering. Dr. Isiel S. Greenstein is administering this study under a contract with the Naval Ocean Systems Center.

Your task as a participant in this study is to determine specific facts from a number of data sets using the window management system. Participation in the study is entirely voluntary. If you choose to participate, you will receive instruction in the use of the software, you will participate in four experimental sessions, and you will be asked to complete a questionnaire regarding the window management approaches used in the test. The experimental sessions will consist of two 50-minute blocks of trials with brief rest breaks between blocks. The entire experiment will require about seven hours to complete. You will receive a voucher in the amount of 30 dollars upon completion of the study, including training, rest breaks and questionnaire administration.

We hope that this experiment will be an interesting experience for you. It is possible that at times you may feel frustrated or stressed. Your performance on the task reflects the difficulty of the task, not your personal abilities or talents.

Please note:

- 1. You have the right to stop participating in the experiment at any time. If you choose to terminate your participation in the experiment, you will receive payment only for the proportion of time you participated.
- 2. You have the right to see your data and to withdraw them from the experiment. If you decide to withdraw your data, please notify the experimenter immediately. Otherwise identification of your particular data will not be possible because the data will be separated from your code number.

Please, name the subject to be informed of the results of this experiment. If other participants want to receive information regarding this study, please indicate your address. This subject's name will be utilized below. If more detailed information is desired, this can be obtained by contacting the Office of the Director, and a full report will be made available to you.

Your participation is greatly appreciated. If you have any questions about the experiment or your rights as a participant, please do not hesitate to ask. We will answer your questions as openly and honestly as is possible without biasing the experimental results. Should you have any additional questions or problems, contact Dr. Melvin Greenstein, Associate Professor at NIH 1642.

Your signature below indicates you have read the above stated rights and will consent to participate. If you include your printed name and address below, a summary of the experimental results will be sent to you.

Signature

Printed Name

Address

City, State, Zip

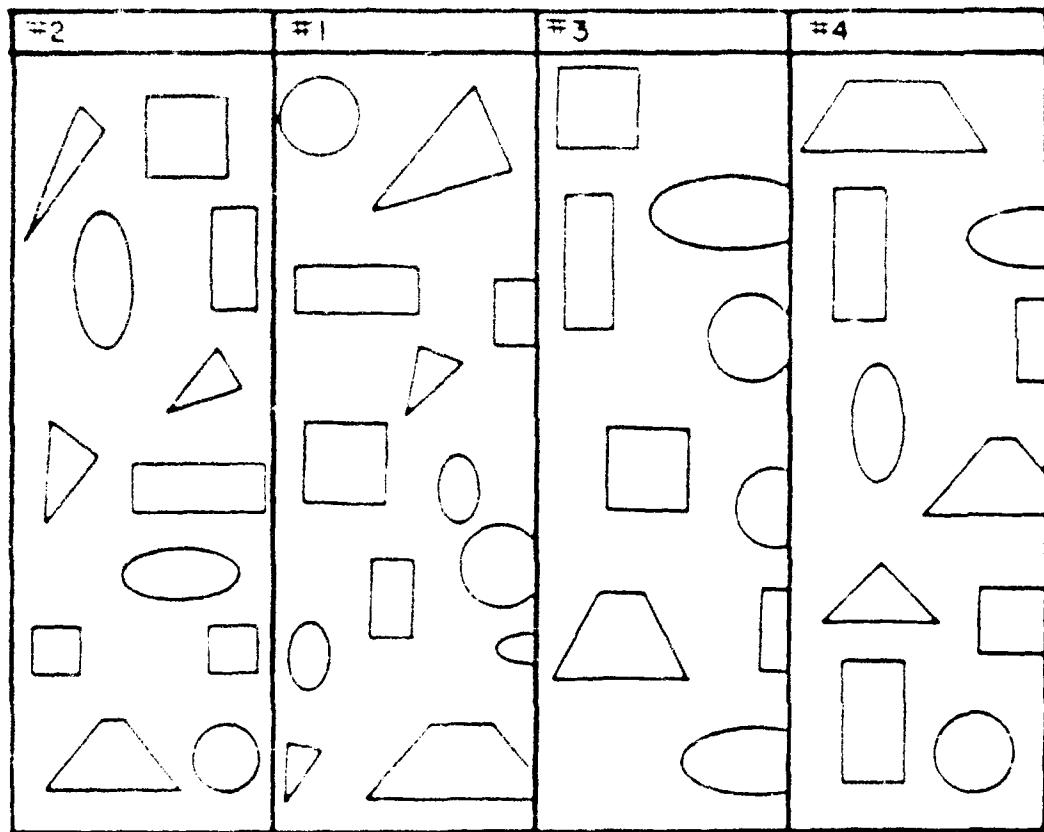
Methodology

Subjects = The Department of Defense. Figure 8-1 shows the subjects' characteristics. We approached our investigation in three phases.

A windowing system allows the user of a computer to simultaneously access and act upon multiple sources of information as if he had multiple video displays connected to one computer. Figure 8-1 shows four windows displayed on the video display at once. Each of these windows contains different information. This study investigates two different techniques commonly used in "tiled" window management systems. The first, termed the localized approach, creates space for a new or enlarged window by taking space from adjacent windows. When the adjacent windows do not have enough space to meet the needs of the new or enlarged window, space is also taken from the next most adjacent window. The second, termed the global approach, creates space for a new or enlarged window by taking equal amounts of space from other windows in the same row of the display. Figures 8-2, 8-3, and 8-4 show the difference between these two approaches.

Design

The hardware configuration will consist of a Tandy 1000 personal computer with keyboard interfaced to a Tandy VM-3 text monitor and an NEC graphics monitor. All textual input and output questions that you are to answer, window management commands that you enter, answers that you enter, and feedback directed to you regarding your answers, will be printed to the text monitor. All data sets containing graphics which are displayed in a graphical windowed environment on the video monitor. Software has been written to estimate task times. Estimated task



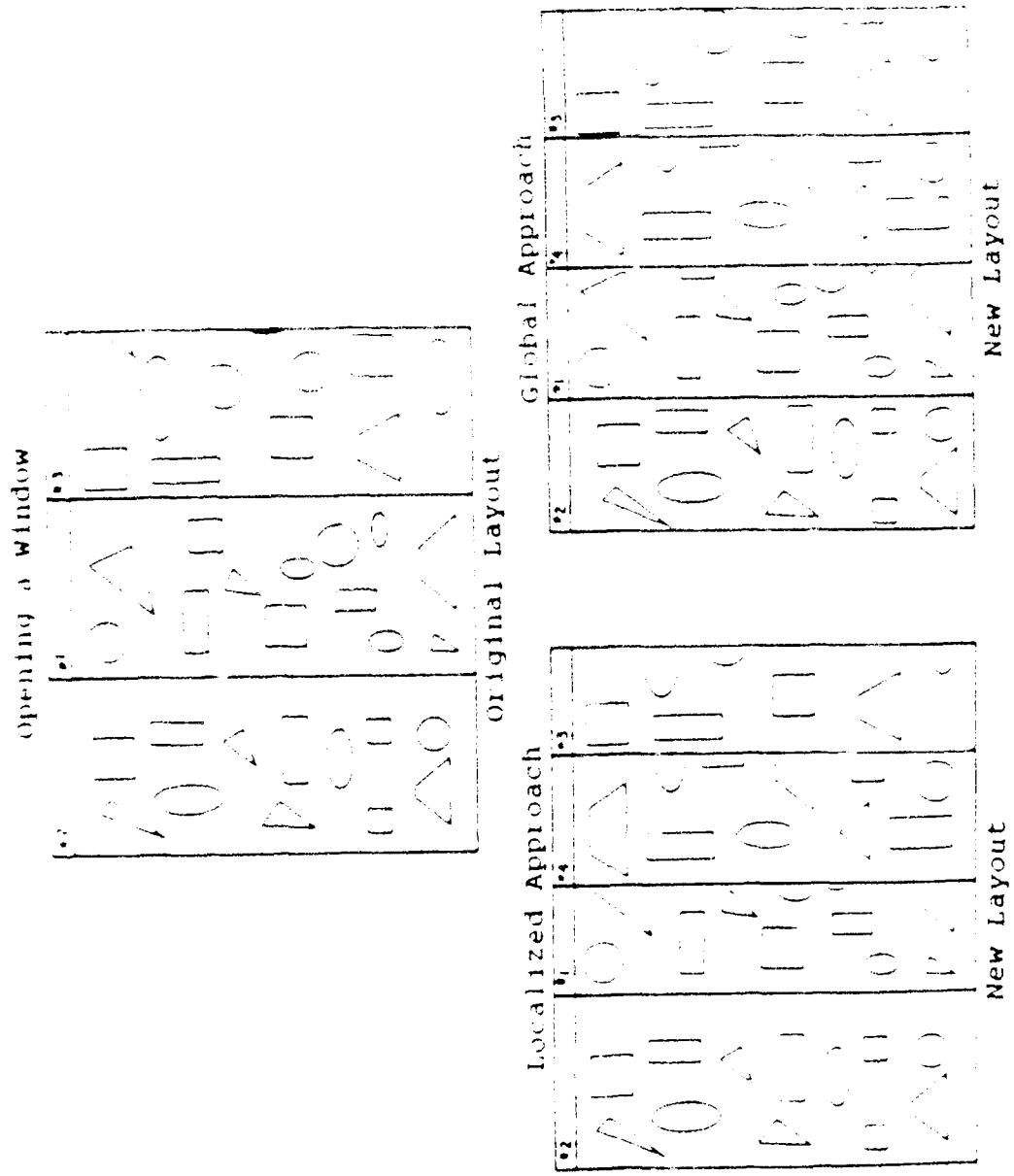


Fig. 11. Comparison of Localized Approach vs. Global Approach in opening a window (a)

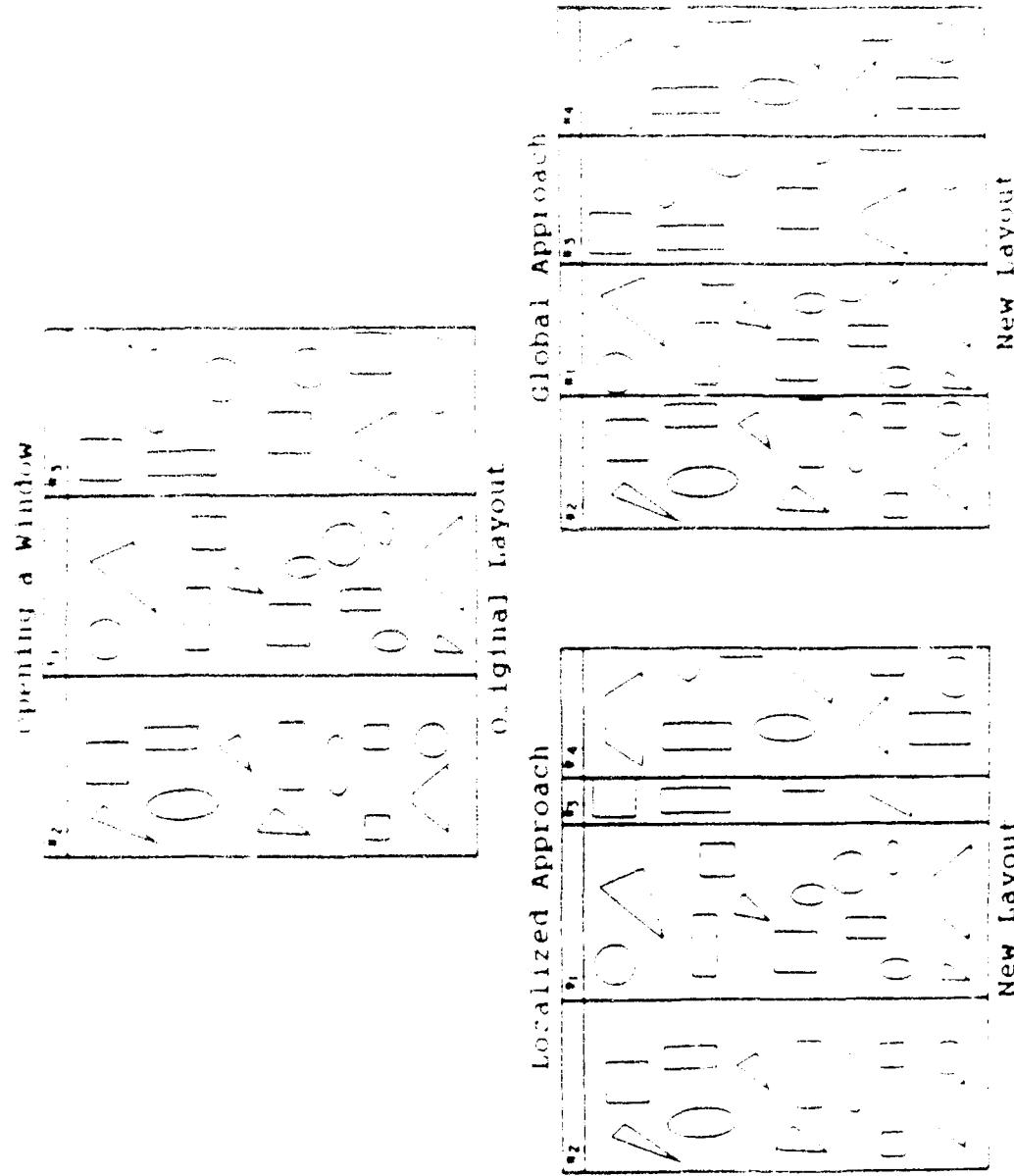
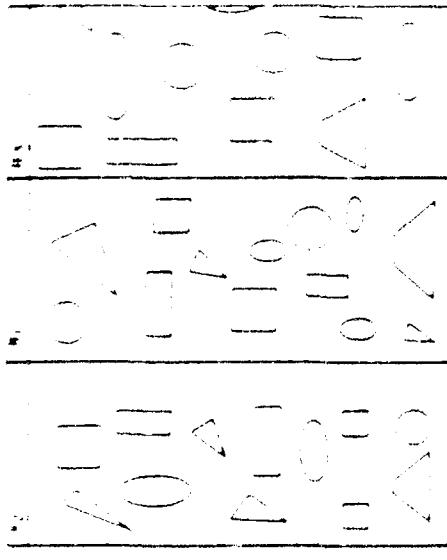


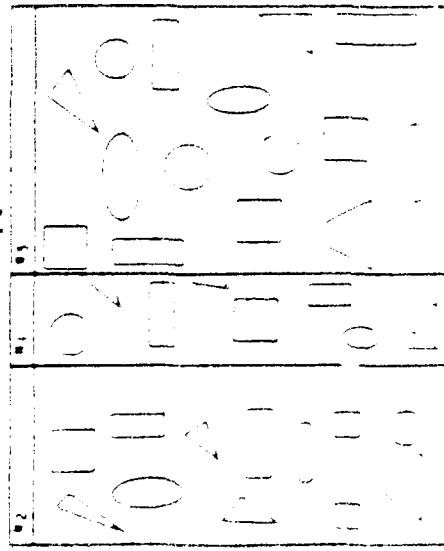
FIGURE 8: Comparison of Localized vs. Global Approach for opening a window [10].

Enlarging a Window



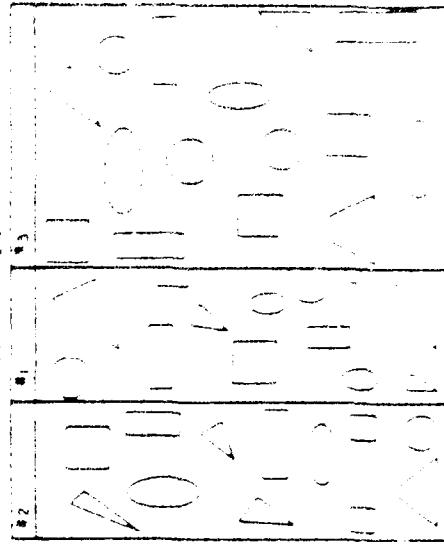
original layout

localized Approach



New Layout

Global Approach



New Layout

Figure 2. Localized Approach vs. Global Approach to Enlarging a Window

The first window will contain a single window which you must move and resize within an existing window when specified. Subsequent windows will contain a series of overlapping windows. You must open and close each window in turn to solve the simplified problem.

Experimental Task

The task environment consists of a number of data sets. Each window has an immediate data set. Data sets are organized in pairs, of two-dimensional objects like circles, ellipses, squares, rectangles, parallelograms, and trapezoids. You are asked to answer some questions about the contents of these data sets using the window management system. Please try to answer these questions as quickly as possible while minimizing incorrect answers.

Measures

Response accuracy, number of window management operations (i.e., openings, enlargings and reopenings), and task completion time will be recorded. Task completion time will be divided into the following:

1. Window Management Time: This includes any time you spend arranging the display screen into a configuration that is suitable to complete the subtask. Screen arrangement time itself will be subdivided into the following components:

Opening Time--this includes time you spend opening windows

Enlarging Time--this includes time you spend enlarging windows

Reopening Time--this includes time you spend reopening windows

2. Task Solution Time: This measures the time you devote to determining the specific tasks requested once you have completed opening and enlarging of the relevant windows.

You will be asked to complete a subjective rating scale just before the completion of each task. A final questionnaire will also be administered at the end of the test.

STUDY DESIGN

The window management system experiments will be divided into two separate studies. Participants will be assigned to one of three groups. All participants will use either the two window management systems or complete the task using eight task blocks.

- 1. localized approach, low data set size variability, and low data set interdependency
- 2. localized approach, low data set size variability, and low data set interdependency
- 3. localized approach, high data set size variability, and low data set interdependency
- 4. localized approach, high data set size variability, and high data set interdependency
- 5. global approach, low data set size variability, and low data set interdependency
- 6. global approach, low data set size variability, and high data set interdependency
- 7. global approach, high data set size variability, and low data set interdependency
- 8. global approach, high data set size variability, and high data set interdependency

The order of these eight blocks will be different for different participants. Each of these eight task blocks requires you to extract information contained in 10 data sets to answer 36 questions.

Before each of the eight task blocks a training session will be provided. Each block of the study will require about fifty minutes to complete. Brief rest breaks will be provided between blocks. The entire experiment will require about seven hours to complete, including administration of instructions, rest breaks, and completion of the rating scale questionnaire.

DATA MENU

QUESTION 1

All the screens have been designed to be used with the keyboard. In window management systems used in office terminals, windows can be defined in two ways. There is no difference between the two approaches to window management in this demonstration. One way is to have a separate window manager used in the terminal; another approach is to have the window management system integrated into the terminal. This demonstration employs the global approach.

READY - please type DEMO and press RETURN

You are asked to enter your code number to start the demo. You must enter the code number that I gave you and hit the ENTER key to start the demo.

You can see that there are a series of small rectangles each containing a different character shown at the top left corner of the left and side display titled "DATA MENU". These small rectangles represent the data sets - at present you cannot see the contents of these data sets. The character inside each small rectangle represents a data set identifier. If you look at the text shown on the right hand side display, you can see the main command menu and question 1. This command menu reminds you how to use the four window management commands available to you. Question 1 asks you to answer the question "What is the total number of rectangles contained in data sets b and d?" Since you cannot see the contents of data sets b and d, you must open windows to these windows sets to answer this question. To open a window for data set b, type the command **OPEN B**. You will see a window appear on the right hand side of the screen. It contains the character "b" which is the first character of the string "OPEN B". The window is a rectangle with a border around it. Type **OPEN D**.

THE NEXT QUESTION IS AS FOLLOWS: "What is the total number of rectangles shown in the right hand side window?". It is the second row of the display. You can see on the left side of the window two small rectangles and one large rectangle. Hit the right arrow key twice. The cursor will move to the middle of the next small rectangle containing character b. Now hit the ENTER key and you will see the following message: "containing character b has become a big rectangle". The other two data sets are shown in this big rectangle, called a window. Hit the right arrow key again. Now move the arrow cursor by hitting the right arrow key twice to the small rectangle containing character c. Hit ENTER. And the small rectangle containing character c will become a big rectangle. Window d. It can be seen in the second row. "window b" is shown in the first row. To answer the question you have to count the total number of rectangles contained in data sets b and d. When you are ready to enter your answer, hit "a" and ENTER. The instruction "GIVE YOUR ANSWER" will then appear on the right hand side display. The answer to this question turns out to be 6. Type "6" and ENTER. You will be given the feedback "This answer is correct, enter any key to continue". If the answer you gave had been incorrect, the feedback would have been "This answer is incorrect. The correct answer is 6". Hit any key to continue. In either case, the system will continue to the next question when you hit any key as instructed. Now hit any key to continue.

The second question is "What is the difference between the total number of circles contained in data set b and the total number of triangles contained in data set r?". Since data set b contains 4 circles and data set r contains 3 circles, the answer is 1. Type "1" and ENTER. You will be given the feedback "This answer is correct, enter any key to continue".

Now hit ENTER. You'll see that a small rectangle has appeared
and then hit the ENTER key again without pressing "fkey". Try it.
Whenever you type an invalid command and hit ENTER, a beep will sound to
indicate that you have made a typing mistake. Now hit 'c' and ENTER
correctly. Use the right direction key to move the arrow cursor to the
small rectangle containing character r and hit ENTER. This time you may
be surprised to see that the small rectangle does not become a big rec-
tangle. Instead an arrow cursor appears at the left boundary of the
first window in the first row (window b). If you look at the right hand
side display, you can see the new instruction "USE THE DIRECTION KEY TO
CHOOSE A POSITION". You haven't done anything wrong. After two windows
have been opened, you have to specify the position for any additional
windows you want to open. You specify a position for the new window by
placing the arrow cursor on the extreme left boundary of a row, the
extreme right boundary of a row, or between two opened windows in a row.
If you hit ENTER now, you will see that window r is placed on the left
hand side of window b and the size of window b is reduced (since window
b gives some of its space to window r). Count the requested graphic
objects and hit 'a' to give your answer.

demo2 (please type demo2 and press ENTER)

Enter your code number and hit the ENTER key to start the demo.

You can see there are already eleven open windows on the display.

In these windows, the contents of some of the data sets are fully visi-
ble, while the contents of other data sets are partly visible. You can
tell whether the contents of a data set are fully visible by looking for
an asterisk (*) at the upper right corner of the window. If there is an
asterisk displayed, the full contents of that data set are visible in

the window. Now look at the question on the right hand side display. The question requires you to count and enter the total number of squares in data set e and g. Since you cannot view the full contents of either data set e or data set g, you will have to enlarge their windows.

To enlarge window e, first hit 'e' for "enlarge" and ENTER. On the right hand side display you can see the new instruction "USE DIRECTION KEYS TO CHOOSE THE WINDOW WHICH YOU WANT TO ENLARGE". If you look at the left hand side display, you can see a cross cursor in the middle of the first window in row one (window k). Hit the right direction key four times. This moves the cross cursor to the middle of window e. Hit the ENTER key. You will now see the new instruction "USE LEFT OR RIGHT DIRECTION KEY TO ENLARGE THE WINDOW" on the right hand side display. The cross cursor automatically moves to the left boundary of window e (since there are no windows on the right hand side of window e, window e can only be enlarged towards the left). Now hit the left direction key ten times. Then hit ENTER. You can see that window e is enlarged, but the contents within it are still not fully displayed, while the size of window o is reduced (since it gave up some of its space to window e). You will have to enlarge window e further (following the same procedure). This time, when you enlarge window e, hit the left direction key five times. You can see from the asterisk that the full contents of data set e are now displayed. Since we have to get information from windows e and g, we will have to enlarge window g as well. Hit 'e' (for "enlarge") and ENTER. Then hit the down direction key once. Now hit the right direction key three times. The cross cursor should now be in the middle of window f. If you hit the ENTER key now, the system will allow you to enlarge window f. Try it. The

cross cursor stays in the middle of window f. Since there are windows on both sides of window f, the system allows you to enlarge window f by moving the cross cursor towards either side. The system now asks you to use the left or right direction key to enlarge window f. Look at the new instruction shown on the right hand side display. But wait! You do not want to enlarge window f! We are supposed to enlarge window g! You can hit the ESC key to respecify the window which you want to enlarge - try it and look at the new instruction on the right hand side display. The new instruction on the right hand side display asks you to "USE DIRECTION KEYS TO CHOOSE THE WINDOW WHICH YOU WANT TO ENLARGE". Hit the down direction key once and hit the right direction key four times. The cross cursor should now be in the middle of window g. Hit ENTER. The cross cursor will stay in the middle of window g. Hit the left direction key 14 times. Then hit ENTER. You can now view the full contents of data set g. But notice that window f has been closed and data set f has been hidden (data set f now appears as a small rectangle in the second row of small rectangles titled "HIDDEN DATA SETS" in the top left corner of the display.) A window will be closed automatically when it is fully covered by another window. This happens when the space left for the window is very small. Now, since you can view the full contents of data sets e and g, you can count the specified graphic objects and answer the question.

We will now proceed to the next question. Question 1+ asks you "What is the total number of parallelograms contained in data sets f and g?" Since a window for data set p hasn't yet been opened, you will have to open it. Follow the procedure to open a window and put window p between windows g and h. Window p will be opened, but you will not be

able to view the full contents of window p. There is no material in the upper right corner of window p. When we opened window p between windows g and h, window g and h had to give up space for the new window. As a result, the sizes of windows g and h were reduced. Window f was so small after giving up space, that it was automatically closed. Data set t now appears as a small rectangle in the second row of small rectangles in the top left corner of the display. Since you cannot view the full contents of window p, you will have to enlarge window p. Use the enlarging procedure to do this. After enlarging window p, you will also have to reopen window f to answer the question. Window f was opened earlier, but was automatically closed when we enlarged window g. To make space for a window that was opened, but later closed, we reopen the window rather than open it. To reopen window f, hit 'r' and then hit ENTER. An arrow cursor is now shown in the first small rectangle of the second row at the top of the display. Hit the right direction key once. The arrow cursor moves to the small rectangle containing character f. Hit the ENTER key. The arrow cursor is now shown at the left boundary of window k. Now hit the right direction key five times. The arrow cursor moves to the right boundary of window row one on the right hand side of window e. Now hit ENTER. The size of window e is reduced to make space for window f. Window f is now reopened. Since you cannot view the full contents of window f, you will have to enlarge it. After you have enlarged window f, count the specified graphic objects, hit a, and give your answer.

ENTER PLEASE TYPE ITEM#1 AND PRESS ENTER

Enter your code number and hit the ENTER key to start the item.

You can see there are already eleven open windows on the display. In these windows, the contents of some of the data sets are fully visible, while the contents of other data sets are partly visible. Now look at the question on the right hand side display. The question requires you to count and enter the total number of squares in data sets e and g. Since you cannot view the full contents of either data set e or data set g, you will have to enlarge their windows.

To enlarge window e, first hit 'e' (for "enlarge") and ENTER. On the right hand side display you can see the new instruction "USE DIRECTION KEYS TO CHOOSE THE WINDOW WHICH YOU WANT TO ENLARGE". If you look at the left hand side display, you can see a cross cursor in the middle of the first window in row one (window k). Hit the right direction key four times. This moves the cross cursor to the middle of window e. Hit the ENTER key. You will now see the new instruction "USE LEFT OR RIGHT DIRECTION KEY TO ENLARGE THE WINDOW" on the right hand side display. The cross cursor automatically moves to the left boundary of window e. Now hit the left direction key ten times. Then hit ENTER. You can see that window e is enlarged, but the contents within it are still not fully displayed, while the sizes of the other windows in the same row are reduced (since they gave up some of their space to window e). You will have to enlarge window e further. This time, when you enlarge window e, hit the left direction key five times. You can see from the asterisk that the full contents of data set e are now displayed. Since we have to get information from windows e and g, we will have to enlarge window g as well. Hit 'e' (for "enlarge") and ENTER.

Then hit the down direction key once. Now hit the right direction key four times. The cross cursor should now be in the middle of window g. Hit ENTER. The cross cursor will stay in the middle of window g. Hit the left direction key 14 times. Then hit ENTER. You can now view the full contents of data set g. But notice that window p has been closed and data set h has been hidden. Now, since you can view the full contents of data sets e and g, you can count the specified alphabetic objects and answer the question.

Question 14 asks you "What is the total number of parallelograms contained in data sets n and p?" Since a window for data set p hasn't yet been opened, you will have to open it. Follow the procedure to open a window and put window p between windows g and h. Window p will be opened, but you will not be able to view the full contents of data set p. When we opened window p in row two, all the other windows in row two had to give up space for the newly opened window. As a result, window h was so small after giving up space, that it was automatically closed. Data set h now appears as a small rectangle in the second row of small rectangles in the top left corner of the display. Since you cannot view the full contents of window p, you will have to enlarge window p. Use the enlarging procedure to do this. After enlarging window p, you will also have to reopen window n to answer the question. To reopen window n, hit 'r' and then hit ENTER. An arrow cursor is now shown in the first small rectangle of the second row at the top of the display. Hit the right direction key once. The arrow cursor moves to the small rectangle containing character n. Hit the ENTER key. The arrow cursor is now shown at the left boundary of window k. Now hit the right direction key five times. The arrow cursor moves to the right boundary of window

row one on the right hand side of window n. Now hit ENTER. The windows at all the other windows in this row are reduced to make space for window n. Window n is now reopened. Since you cannot view the full contents of window n, you will have to enlarge it. After you have enlarged window n, count the specified graphic objects, hit 'a' and give your answer.

NOTES:

1. You can use the backspace key to correct your input before you hit the ENTER key
2. The ESC key can be very useful when you type a command and enter it, but then change your mind. Hitting the ESC key at this point will "undo" the last command you entered and bring the system back to where it was before you entered that command. But there are four exceptions about the ESC key
 - a. You cannot "undo" the answer you gave.
 - b. You cannot "undo" a window that has been opened
 - c. You cannot "undo" a window that has been enlarged.
 - d. You cannot "undo" a window that has been reopened.

Appendix D

Test Item 1

a. When the localized approach to window management easy to learn

1 extremely easy 2 3 4 5
moderately easy very difficult

b. When the global approach to window management easy to learn

1 extremely easy 2 3 4 5
moderately easy very difficult

c. localized approach, low data set size variability, and low data set interdependency.

1a When the data set sizes are low variability, the data sets are low interdependency, and the screen hasn't yet been filled with data sets, how satisfied are you with the localized approach to window management?

1 extremely satisfied 2 3 just satisfied 4 5
not satisfied at all

1b When the data set sizes are low variability, the data sets are low interdependency, and the screen has been filled with data sets, how satisfied are you with the localized approach to window management?

1 extremely satisfied 2 3 just satisfied 4 5
not satisfied at all

1c When the data set sizes are low variability, the data sets are low interdependency, how satisfied are you with the localized approach to window management?

1 extremely satisfied 2 3 just satisfied 4 5
not satisfied at all

d. localized approach, low data set size variability, and high data set interdependency.

- a. When the data set sizes are low variability, the data sets are low interdependency, and the screen hasn't yet been filled with data sets, how satisfied are you with the localized approach to window management?

1	2	3	4	5
extremely unsatisfied		just satisfied		not satisfied at all

- b. When the data set sizes are low variability, the data sets are high interdependency, and the screen has been filled with data sets, how satisfied are you with the localized approach to window management?

1	2	3	4	5
extremely unsatisfied		just satisfied		not satisfied at all

- c. When the data set sizes are low variability, the data sets are high interdependency, how satisfied are you with the localized approach to window management?

1	2	3	4	5
extremely satisfied		just satisfied		not satisfied at all

d. localized approach - high data set size variability, and low data set interdependency:

- a. When the data set sizes are high variability, the data sets are low interdependency, and the screen hasn't yet been filled with data sets, how satisfied are you with the localized approach to window management?

1	2	3	4	5
extremely satisfied		just satisfied		not satisfied at all

b. When the data set sizes are high variability, the data sets are low interdependency, and the screen has been filled with data sets, how satisfied are you with the localized approach to window management?

1	2	3	4	5
extremely satisfied		just satisfied		not satisfied at all

c. When the data set sizes are high variability, the data sets are low interdependency, how satisfied are you with the localized approach to window management?

1	2	3	4	5
extreme- ly unsatisfied		just satisfied		not satisfied at all

a When the data set sizes are high variability, the data sets are low interdependency.

b When the data set sizes are high variability, the data sets are high interdependency, and the screen hasn't yet been filled with data sets, how satisfied are you with the localized approach to window management?

1	2	3	4	5
extremely satisfied		just satisfied		not satisfied at all

c When the data set sizes are high variability, the data sets are high interdependency, and the screen has been filled with data sets, how satisfied are you with the localized approach to window management?

1	2	3	4	5
extremely satisfied		just satisfied		not satisfied at all

d When the data set sizes are high variability, the data sets are high interdependency, how satisfied are you with the localized approach to window management?

1	2	3	4	5
extremely satisfied		just satisfied		not satisfied at all

global approach (low data set size variability, and low data set interdependency)

a When the data set sizes are low variability, the data sets are low interdependency, and the screen hasn't yet been filled with data sets, how satisfied are you with the global approach to window management?

1	2	3	4	5
extremely satisfied		just satisfied		not satisfied at all

b When the data set sizes are low variability, the data sets are low interdependency, and the screen has been filled with data sets, how satisfied are you with the global approach to window management?

1	2	3	4	5
extremely satisfied		just satisfied		not satisfied at all

7c When the data set sizes are low variability, the data sets are high interdependency, how satisfied are you with the global approach to window management?

1 2 3 4 5
extremely satisfied just satisfied not satisfied at all

8 (global approach, low data set size variability, and high data set interdependency)

9a When the data set sizes are low variability, the data sets are high interdependency, and the screen hasn't yet been filled with data sets, how satisfied are you with the global approach to window management?

1 2 3 4 5
extremely satisfied just satisfied not satisfied at all

9b When the data set sizes are low variability, the data sets are high interdependency, and the screen has been filled with data sets, how satisfied are you with the global approach to window management?

1 2 3 4 5
extremely satisfied just satisfied not satisfied at all

10c When the data set sizes are low variability, the data sets are high interdependency, how satisfied are you with the global approach to window management?

1 2 3 4 5
extremely satisfied just satisfied not satisfied at all

11 (global approach, high data set size variability, and low data set interdependency)

12a When the data set sizes are high variability, the data sets are low interdependency, and the screen hasn't yet been filled with data sets, how satisfied are you with the global approach to window management?

1 2 3 4 5
extremely satisfied just satisfied not satisfied at all

4. When the data set sizes are high variability, the data sets are high interdependency, and the screen hasn't yet been filled with data sets, how satisfied are you with the global approach to window management?

1	2	3	4	5
extremely satisfied		just satisfied		not satisfied at all

4a. When the data set sizes are high variability, the data sets are low interdependency, how satisfied are you with the global approach to window management?

1	2	3	4	5
extremely satisfied		just satisfied		not satisfied at all

5. global approach, high data set size variability, and high data set interdependency

5a. When the data set sizes are high variability, the data sets are high interdependency, and the screen hasn't yet been filled with data sets, how satisfied are you with the global approach to window management?

1	2	3	4	5
extremely satisfied		just satisfied		not satisfied at all

5b. When the data set sizes are high variability, the data sets are high interdependency, and the screen has been filled with data sets, how satisfied are you with the global approach to window management?

1	2	3	4	5
extremely satisfied		just satisfied		not satisfied at all

5c. When the data set sizes are high variability, the data sets are high interdependency, how satisfied are you with the global approach to window management?

1	2	3	4	5
extremely satisfied		just satisfied		not satisfied at all

1. How satisfied are you with the localized approach to window management?

extremely	just	not satisfied
satisfied	satisfied	at all

2. How satisfied are you with the global approach to window management?

extremely	just	not satisfied
satisfied	satisfied	at all

Rank these two window management approaches in order of preference with 1 being more preferred and 2 less preferred.

Localized approach

Global approach

100 Order Independent Data

Question 1:

What is the difference between the total number of ellipses and rectangles contained in data set g and the total number of ellipses and rectangles contained in data set k?

Question 2:

What is the total number of circles and squares contained in data sets g and k?

Question 3:

What is the difference between the total number of ellipses and circles contained in data set s and the total number of trapezoids and parallelograms contained in data set g?

Question 4:

What is the total number of rectangles and trapezoids contained in data sets g and k?

Question 5:

What is the difference between the total number of squares and rectangles contained in data set s and the total number of trapezoids and ellipses contained in data set e?

Question 6:

What is the difference between the total number of circles and parallelograms contained in data set g and the total number of circles and parallelograms contained in data set s?

Question 7:

What is the difference between the total number of ellipses and rectangles contained in data set i and the total number of circles and parallelograms contained in data set e?

Question 8:

What is the total number of circles and squares contained in data sets g and si?

Question 9:

What is the difference between the total number of ellipses and squares contained in data set k and the total number of ellipses and squares contained in data set si?

Question 10:

What is the difference between the total number of trapezoids and parallelograms contained in data set j and the total number of ellipses and rectangles contained in data set e?

Question 11:

What is the difference between the total number of ellipses and rectangles contained in data set b and the total number of ellipses and rectangles contained in data set s?

Question 12:

What is the total number of rectangles and parallelograms contained in data sets g and t?

Question 13:

What is the difference between the total number of rectangles and trapezoids contained in data set k and the total number of rectangles and trapezoids contained in data set p?

Question 14:

What is the total number of parallelograms and squares contained in data sets e and j?

Question 15:

What is the difference between the total number of squares and circles contained in data set q and the total number of trapezoids and parallelograms contained in data set b?

Question 16:

What is the total number of circles and trapezoids contained in data sets g and s?

Question 17:

What is the difference between the total number of ellipses and squares contained in data set c and the total number of trapezoids and rectangles contained in data set t?

Question 18:

What is the difference between the total number of ellipses and squares contained in data set k and the total number of ellipses and squares contained in data set p?

Question 19:

What is the difference between the total number of ellipses and rectangles contained in data set h and the total number of ellipses and rectangles contained in data set n?

Question 20:

What is the total number of circles and squares contained in data sets e and j?

Question 21:

What is the difference between the total number of ellipses and circles contained in data set q and the total number of trapezoids and parallelograms contained in data set d?

Question 22:

What is the total number of rectangles and trapezoids contained in data sets b and s?

Question 23:

What is the difference between the total number of squares and rectangles contained in data set i and the total number of trapezoids and ellipses contained in data set g?

Question 24:

What is the difference between the total number of circles and parallelograms contained in data set c and the total number of circles and parallelograms contained in data set t?

Question 25:

What is the difference between the total number of ellipses and rectangles contained in data set l and the total number of circles and parallelograms contained in data set p?

Question 26:

What is the total number of circles and squares contained in data sets h and k?

Question 27:

What is the difference between the total number of ellipses and squares contained in data set a and the total number of ellipses and squares contained in data set n?

Question 28:

What is the difference between the total number of trapezoids and parallelograms contained in data set j and the total number of ellipses and rectangles contained in data set e?

Question 29:

What is the difference between the total number of ellipses and rectangles contained in data set m and the total number of ellipses and rectangles contained in data set q?

Question 30:

What is the total number of rectangles and parallelograms contained in data sets d and s?

Question 31:

What is the difference between the total number of rectangles and trapezoids contained in data set i and the total number of rectangles and trapezoids contained in data set r?

Question 32:

What is the total number of parallelograms and squares contained in data sets b and g?

Question 33:

What is the difference between the total number of squares and circles contained in data set c and the total number of trapezoids and parallelograms contained in data set f?

Question 34:

What is the total number of circles and trapezoids contained in data sets l and m?

Question 35:

What is the difference between the total number of ellipses and squares contained in data set p and the total number of trapezoids and rectangles contained in data set o?

Question 36:

What is the difference between the total number of ellipses and squares contained in data set h and the total number of ellipses and squares contained in data set k?

High Interdependency Task

Question 1:

What is the difference between the total number of ellipses and rectangles contained in data set e and the total number of ellipses and rectangles contained in data set j?

Question 2:

What is the difference between the total number of circles and squares contained in data set e and the total number of circles and squares contained in data set j?

Question 3:

What is the difference between the total number of ellipses and circles contained in data set p and the total number of trapezoids contained in data sets e and j?

Question 4:

What is the difference between the total number of rectangles and squares contained in data set p and the total number of parallelograms contained in data sets e and j?

Question 5:

What is the sum of the number of trapezoids contained in data set p, ellipses contained in data set j, circles contained in data set t, and rectangles contained in data set e?

Question 6:

What is the total number of parallelograms contained in data sets e, j, p, and t?

Question 7:

What is the sum of the number of ellipses contained in data set p, rectangles contained in data set j, squares contained in data set q, and trapezoids contained in data set t?

Question 8:

What is the difference between the total number of circles contained in data sets p and t and the total number of squares contained in data sets e and j?

Question 9:

What is the total number of rectangles contained in data sets c, p, i, and t?

Question 10:

What is the sum of the number of squares contained in data set p, parallelograms contained in data set e, trapezoids contained in data set q, and ellipses contained in data set j?

Question 11:

What is the total number of circles contained in data sets c, e, i, and q?

Question 12:

What is the difference between the total number of trapezoids contained in data sets p and q and the total number of rectangles contained in data sets j and t?

Question 13:

What is the total number of squares contained in data sets c, e, i, and k?

Question 14:

What is the difference between the total number of rectangles contained in data sets p and q and the total number of trapezoids contained in data sets j and t?

Question 15:

What is the sum of the number of circles contained in data set e, squares contained in data set b, ellipses contained in data set c, and parallelograms contained in data set i?

Question 16:

What is the difference between the total number of trapezoids contained in data sets k and p and the total number of ellipses contained in data sets j and t?

Question 17:

What is the sum of the number of ellipses contained in data set e, circles contained in data set i, parallelograms contained in data set q, and squares contained in data set o?

Question 18:

What is the total number of circles contained in data sets b, c, j, and t?

Question 19:

What is the difference between the total number of ellipses and rectangles contained in data set f and the total number of ellipses and rectangles contained in data set r?

Question 20:

What is the difference between the total number of circles and squares contained in data set k and the total number of circles and squares contained in data set p?

Question 21:

What is the difference between the total number of ellipses and circles contained in data set a and the total number of trapezoids contained in data sets e and i?

Question 22:

What is the difference between the total number of rectangles and squares contained in data set q and the total number of parallelograms contained in data sets j and o?

Question 23:

What is the sum of the number of trapezoids contained in data set c, ellipses contained in data set t, circles contained in data set h, and rectangles contained in data set b?

Question 24:

What is the total number of parallelograms contained in data sets f, k, p, and r?

Question 25:

What is the sum of the number of ellipses contained in data set e, rectangles contained in data set m, squares contained in data set a, and trapezoids contained in data set i?

Question 26:

What is the difference between the total number of circles contained in data sets j and q and the total number of squares contained in data sets b and o?

Question 27:

What is the total number of rectangles contained in data sets c, d, n, and t?

Question 28:

What is the sum of the number of squares contained in data set f, parallelograms contained in data set k, trapezoids contained in data set r, and ellipses contained in data set p?

Question 29:

What is the total number of circles contained in data sets a, i, m, and n?

Question 1:

What is the difference between the total number of trapezoids contained in data sets b and e and the total number of rectangles contained in data sets c and d?

Question 2:

What is the total number of squares contained in data sets b, f, g, s, and t?

Question 3:

What is the difference between the total number of rectangles contained in data sets f and k and the total number of trapezoids contained in data sets c and d?

Question 4:

What is the sum of the number of squares contained in data set g, ellipses contained in data set m, circles contained in data set r, and parallelograms contained in data set p?

Question 5:

What is the difference between the total number of trapezoids contained in data sets a and b and the total number of ellipses contained in data sets f and h?

Question 6:

What is the sum of the number of ellipses contained in data set z, circles contained in data set l, parallelograms contained in data set e, and squares contained in data set o?

Question 7:

What is the total number of circles contained in data sets h, j, s, and t?

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